SDMX Technical Working Group

VTL Task Force

VTL - version 2.0
(Validation & Transformation Language)

Part 2 - Reference Manual

July 2018
The Task force for the Validation and Transformation Language (VTL), created in 2012-2013 under the initiative of the SDMX Secretariat, is pleased to present the draft version of VTL 2.0.

The SDMX Secretariat launched the VTL work at the end of 2012, moving on from the consideration that SDMX already had a package for transformations and expressions in its information model, while a specific implementation language was missing. To make this framework operational, a standard language for defining validation and transformation rules (operators, their syntax and semantics) had to be adopted, while appropriate SDMX formats for storing and exchanging rules, and web services to retrieve them, had to be designed. The present VTL 2.0 package is only concerned with the first element, i.e., a formal definition of each operator, together with a general description of VTL, its core assumptions and the information model it is based on.

The VTL task force was set up early in 2013, composed of members of SDMX, DDI and GSIM communities and the work started in summer 2013. The intention was to provide a language usable by statisticians to express logical validation rules and transformations on data, described as either dimensional tables or unit-record data. The assumption is that this logical formalization of validation and transformation rules could be converted into specific programming languages for execution (SAS, R, Java, SQL, etc.), and would provide at the same time, a "neutral" business-level expression of the processing taking place, against which various implementations can be mapped. Experience with existing examples suggests that this goal would be attainable.

An important point that emerged is that several standards are interested in such a kind of language. However, each standard operates on its model artefacts and produces artefacts within the same model (property of closure). To cope with this, VTL has been built upon a very basic information model (VTL IM), taking the common parts of GSIM, SDMX and DDI, mainly using artefacts from GSIM 1.1, somewhat simplified and with some additional detail. In this way, existing standards (GSIM, SDMX, DDI, others) would be allowed to adopt VTL by mapping their information model against the VTL IM. Therefore, although a work-product of SDMX, the VTL language in itself is independent of SDMX and will be usable with other standards as well. Thanks to the possibility of being mapped with the basic part of the IM of other standards, the VTL IM also makes it possible to collect and manage the basic definitions of data represented in different standards.

For the reason described above, the VTL specifications are designed at logical level, independently of any other standard, including SDMX. The VTL specifications, therefore, are self-standing and can be implemented either on their own or by other standards (including SDMX). In particular, the work for the SDMX implementation of VTL is going in parallel with the work for designing this VTL version, and will entail a future update of the SDMX documentation.

The first public consultation on VTL (version 1.0) was held in 2014. Many comments were incorporated in the VTL 1.0 version, published in March 2015. Other suggestions for improving the language, received afterwards, fed the discussion for building the draft version 1.1, which contained many new features, was completed in the second half of 2016 and provided for public consultation until the beginning of 2017.

The high number and wide impact of comments and suggestions induced a high workload on the VTL TF, which agreed to proceed in two steps for the publication of the final documentation, taking also into consideration that some first VTL implementation initiatives had already been launched. The first step, the current one, is dedicated to fixing some high-priority features and making them as much stable as possible. A second step, scheduled for the next period, is aimed at acknowledging and fixing other features considered of minor impact and priority, which will be added hopefully without affecting neither the features already published in this documentation, nor the possible relevant implementations. Moreover, taking into account the number of very important new features that have been introduced in this version in respect to the VTL 1.0, it was agreed that the current VTL version should be considered as a major one and thus named VTL 2.0.

The VTL 2.0 package contains the general VTL specifications, independently of the possible implementations of other standards; in its final release, it will include:

a) Part 1 – the user manual, highlighting the main characteristics of VTL, its core assumptions and the information model the language is based on;

b) Part 2 – the reference manual, containing the full library of operators ordered by category, including examples; this version will support more validation and compilation needs compared to VTL 1.0.

c) eBNF notation (extended Backus-Naur Form) which is the technical notation to be used as a test bed for all the examples.

The present document is the part 2.
The latest version of VTL is freely available online at https://sdmx.org/?page_id=5096

Acknowledgements

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Feedback and suggestions for improvement are encouraged and should be sent to the SDMX Technical Working Group (twg@sdmx.org).
# Table of contents

- Foreword .................................................................................................................. 2
- Table of contents ....................................................................................................... 4
- Introduction .................................................................................................................. 8
- Overview of the language and conventions ................................................................... 9
  - Introduction ............................................................................................................... 9
  - Conventions for writing VTL Transformations .......................................................... 10
  - Typographical conventions ....................................................................................... 11
  - Abbreviations for the names of the artefacts ............................................................. 12
  - Conventions for describing the operators’ syntax ...................................................... 12
  - Description of the data types of operands and result .............................................. 14
  - VTL-ML Operators .................................................................................................. 15
  - VTL-ML - Evaluation order of the Operators ............................................................. 27
  - Description of VTL Operators ................................................................................... 27
- VTL-DL - Rulesets ...................................................................................................... 29
  - define datapoint ruleset ............................................................................................ 29
  - define hierarchical ruleset ......................................................................................... 31
- VTL-DL – User Defined Operators ............................................................................. 39
  - define operator ......................................................................................................... 39
  - Data type syntax ....................................................................................................... 40
- VTL-ML - Typical behaviours of the ML Operators ...................................................... 42
  - Typical behaviour of most ML Operators ................................................................ 42
  - Operators applicable on one Scalar Value or Data Set or Data Set Component ........ 42
  - Operators applicable on two Scalar Values or Data Sets or Data Set Components ...... 43
  - Operators applicable on more than two Scalar Values or Data Set Components ....... 45
  - Behaviour of Boolean operators .............................................................................. 45
  - Behaviour of Set operators ...................................................................................... 46
  - Behaviour of Time operators .................................................................................... 46
  - Operators changing the data type ............................................................................ 47
  - Type Conversion and Formatting Mask .................................................................... 48
    - The Numbers Formatting Mask ............................................................................... 48
    - The Time Formatting Mask .................................................................................... 48
  - Attribute propagation ............................................................................................... 51
- VTL-ML - General purpose operators ......................................................................... 53
Parentheses: ( ) ................................................................. 53
Persistent assignment: <= ..................................................... 53
Non-persistent assignment: := ............................................... 55
Membership: # .................................................................. 56
User-defined operator call ....................................................... 57
Evaluation of an external routine: eval ....................................... 58
Type conversion: cast .............................................................. 59
VTL-ML - Join operators ................................................................ 64
Join: inner_join, left_join, full_join, cross_join.......................... 64
VTL-ML - String operators ......................................................... 73
String concatenation: || ............................................................ 73
Whitespace removal: trim, rtrim, ltrim ....................................... 74
Character case conversion: upper/lower ...................................... 75
Sub-string extraction: substr ...................................................... 76
String pattern replacement: replace ............................................ 78
String pattern location: instr ...................................................... 79
String length: length ............................................................... 81
VTL-ML - Numeric operators ....................................................... 83
Unary plus: + ....................................................................... 83
Unary minus: - ..................................................................... 84
Addition: + ....................................................................... 85
Subtraction: - ..................................................................... 87
Multiplication: * .................................................................. 88
Division: / ......................................................................... 90
Modulo: mod ..................................................................... 91
Rounding: round ................................................................... 93
Truncation: trunc .................................................................. 95
Ceiling: ceil ....................................................................... 97
Floor: floor ...................................................................... 98
Absolute value: abs ............................................................... 99
Exponential: exp ................................................................. 100
Natural logarithm: ln ............................................................. 101
Power: power ..................................................................... 103
Logarithm: log .................................................................... 104
Square root: sqrt .................................................................. 105
VTL-ML - Comparison operators......................................................................................... 107
Equal to:    = ........................................................................................................... 107
Not equal to:  <> ............................................................................................................ 108
Greater than:  >     >= ................................................................................................. 109
Less than: <     <= ......................................................................................................... 111
Between: between ........................................................................................................ 112
Element of: in / not_in .................................................................................................... 114
match_characters match_characters ............................................................................. 116
Isnull: isnull.................................................................................................................. 117
Exists in: exists_in ......................................................................................................... 118
VTL-ML - Boolean operators................................................................................................. 121
Logical conjunction:    and ............................................................................................. 121
Logical disjunction:    or ................................................................................................. 122
Exclusive disjunction: xor ............................................................................................... 124
Logical negation: not ...................................................................................................... 126
VTL-ML - Time operators..................................................................................................... 128
Period indicator:    period_indicator .............................................................................. 128
Fill time series:    fill_time_series .............................................................................. 129
Flow to stock: flow_to_stock ......................................................................................... 135
Stock to flow:    stock_to_flow ...................................................................................... 138
Time shift: timeshift ....................................................................................................... 141
Time aggregation:    time_agg ......................................................................................... 145
Actual time:    current_date ............................................................................................ 146
VTL-ML - Set operators..................................................................................................... 148
Union:    union ............................................................................................................... 148
Intersection:    intersect .............................................................................................. 149
Set difference:    setdiff .............................................................................................. 150
Symmetric difference: symdiff ...................................................................................... 152
VTL-ML - Hierarchical aggregation .................................................................................. 154
Hierarchical roll-up: hierarchy ...................................................................................... 154
VTL-ML - Aggregate and Analytic operators .................................................................... 158
Aggregate invocation ....................................................................................................... 159
Analytic invocation ......................................................................................................... 162
Counting the number of data points: count .................................................................... 165
Minimum value:    min ..................................................................................................... 166
Maximum value : `max` ................................................................. 167
Median value : `median` ............................................................ 168
Sum : `sum` ............................................................................. 169
Average value : `avg` ................................................................. 171
Population standard deviation : `stddev_pop` ................................ 172
Sample standard deviation : `stddev_samp` .................................. 173
Population variance : `var_pop` .................................................. 174
Sample variance : `var_samp` ...................................................... 175
First value : `first_value` .......................................................... 176
Last value : `last_value` ............................................................. 177
Lag : `lag` ................................................................................ 179
lead : `lead` ............................................................................. 180
Rank : `rank` ............................................................................ 181
Ratio to report : `ratio_to_report` ................................................ 183

**VTL-ML - Data validation operators** ........................................ 185
check_datapoint ....................................................................... 185
check_hierarchy ....................................................................... 187
check ....................................................................................... 191

**VTL-ML - Conditional operators** ............................................ 194
if-then-else : `if` ...................................................................... 194
Nvl : `nvl` ................................................................................. 196

**VTL-ML - Clause operators** .................................................... 198
Filtering Data Points : `filter` .................................................... 198
Calculation of a Component : `calc` ......................................... 199
Aggregation : `aggr` .................................................................. 200
Maintaining Components : `keep` ............................................. 203
Removal of Components : `drop` .............................................. 204
Change of Component name : `rename` .................................... 205
Pivoting : `pivot` .................................................................... 206
Unpivoting : `unpivot` .............................................................. 207
Subspace : `sub` ..................................................................... 209
This document is the Reference Manual of the Validation and Transformation Language (also known as 'VTL') version 2.0.

The VTL 2.0 library of the Operators is described hereinafter.

VTL 2.0 consists of two parts: the VTL Definition Language (VTL-DL) and the VTL Manipulation Language (VTL-ML).

This manual describes the operators of VTL 2.0 in detail (both VTL-DL and VTL-ML) and is organized as follows.

First, in the following Chapter “Overview of the language and conventions”, the general principles of VTL are summarized, the main conventions used in this manual are presented and the operators of the VTL-DL and VTL-ML are listed. For the operators of the VTL-ML, a table that summarizes the “Evaluation Order” (i.e., the precedence rules for the evaluation of the VTL-ML operators) is also given.

The following two Chapters illustrate the operators of VTL-DL, specifically for:

- the definition of rulesets and their rules, which can be invoked with appropriate VTL-ML operators (e.g., to check the compatibility of Data Point values ...);
- the definition of custom operators/functions of the VTL-ML, meant to enrich the capabilities of the VTL-ML standard library of operators.

The illustration of VTL-ML begins with the explanation of the common behaviour of some classes of relevant VTL-ML operators, towards a good understanding of general language characteristics, which we factor out and do not repeat for each operator, for the sake of compactness.

The remainder of the document illustrates each single operator of the VTL-ML and is structured in chapters, one for each category of Operators (e.g., general purpose, string, numeric ...). For each Operator, there is a specific section illustrating the syntax, the semantics and giving some examples.
Overview of the language and conventions

Introduction

The Validation and Transformation Language is aimed at defining Transformations of the artefacts of the VTL Information Model, as more extensively explained in the User Manual.

A Transformation consists of a statement which assigns the outcome of the evaluation of an expression to an Artefact of the IM. The operands of the expression are IM Artefacts as well. A Transformation is made of the following components:

- A left-hand side, which specifies the Artefact which the outcome of the expression is assigned to (this is the result of the Transformation);
- An assignment operator, which specifies also the persistency of the left hand side. The assignment operators are two, the first one for the persistent assignment (\(<\)-) and the other one for the non-persistent assignment (\(\Rightarrow\)).
- A right-hand side, which is the expression to be evaluated, whose inputs are the operands of the Transformation. An expression consists in the invocation of VTL Operators in a certain order. When an Operator is invoked, for each input Parameter, an actual argument (operand) is passed to the Operator, which returns an actual argument for the output Parameter. In the right hand side (the expression), the Operators can be nested (the output of an Operator invocation can be input of the invocation of another Operator). All the intermediate results in an expression are non-persistent.

Examples of Transformations are:

\[
\text{DS}_{np} := (\text{DS}_1 - \text{DS}_2) \times 2 ; \\
\text{DS}_p \leftarrow \text{if DS}_{np} \geq 0 \text{ then DS}_{np} \text{ else DS}_1 ;
\]

\(\text{DS}_1\) and \(\text{DS}_2\) are input Data Sets, \(\text{DS}_{np}\) is a non persistent result, \(\text{DS}_p\) is a persistent result, the invoked operators (apart the mentioned assignments) are the subtraction (-) the multiplication (*) the choice (if...then...else), the greater or equal comparison (\(\geq\)) and the parentheses that control the order of the operators' invocations.

Like in the example above, Transformations can interact one another through their operands and results; in fact the result of a Transformation can be operand of one or more other Transformations. The interacting Transformations form a graph that is oriented and must be acyclic to ensure the overall consistency, moreover a given Artefact cannot be result of more than one Transformation (the consistency rules are better explained in the User Manual, see VTL Information Model / Generic Model for Transformations / Transformations consistency). In this regard, VTL Transformations have a strict analogy with the formulas defined in the cells of the spreadsheets.

A set of more interacting Transformations is usually needed to perform a meaningful and self-consistent task like for example the validation of one or more Data Sets. The smaller set of Transformations to be executed in the same run is called Transformation Scheme and can be considered as a VTL program.

Not necessarily Transformations need to be written in sequence like a classical software program, in fact they are associated to the Artefacts they calculate, like it happens in the spreadsheets (each spreadsheet’s formula is associated to the cell it calculates).

Nothing prevents, however, from writing the Transformations in sequence, taking into account that not necessarily the Transformations are performed in the same order as they are written, because the order of execution depends on their input-output relationships (a Transformation which calculates a result that is operand of other Transformations must be executed first). For example, if the two Transformations of the example above were written in the reverse order:

\[
\begin{align*}
(i) & \quad \text{DS}_p \leftarrow \text{if DS}_{np} \geq 0 \text{ then DS}_{np} \text{ else DS}_1 ; \\
(ii) & \quad \text{DS}_{np} := (\text{DS}_1 - \text{DS}_2) \times 2 ;
\end{align*}
\]
All the same the Transformation (ii) would be executed first, because it calculates the Data Set DS_np which is an operand of the Transformation (i).

When Transformations are written in sequence, a semicolon (;) is used to denote the end of a Transformation and the beginning of the following one.

Conventions for writing VTL Transformations

When more Transformations are written in a text, the following conventions apply.

**Transformations:**

- A Transformation can be written in one or more lines, therefore the end of a line does not denote the end of a Transformation.
- The end of a Transformation is denoted by a semicolon (;).

**Comments:**

Comments can be inserted within VTL Transformations using the following syntaxes.

- A multi-line comment is embedded between /* and */ and, obviously, can span over several lines:
  
  /* multi-line comment text */

- A single-line comment follows the symbol // up to the next end of line:
  
  // text of a comment on a single line

- A sequence of spaces, tabs, end-of-line characters or comments is considered as a single space.
- The characters /*, */ // and the whitespaces can be part of a string literal (within double quotes) but in such a case they are part of the string characters and do not have any special meaning.

Examples of valid comments:

**Example 1:**

/* this is a multi-line Comment */

**Example 2:**

// this is single-line comment

**Example 3:**

DS_r <- /* A is a dataset */ A + /* B is a dataset */ B;
(for the VTL this statement is the Transformation DS_r <- A + B; )

**Example 4:**

DS_r := DS_1 * DS_2; // my comment
(for the VTL this statement is the Transformation DS_r := DS_1 * DS_2; )
Typographical conventions

The Reference Manual (this manual) uses the normal font Cambria for the text and the other following typographical conventions:

<table>
<thead>
<tr>
<th>Convention</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Italics Cambria</strong></td>
<td>Basic scalar data types <em>(in the text)</em></td>
</tr>
<tr>
<td></td>
<td>e.g. “...must have one Identifier of type time_period. If the Data Set...”</td>
</tr>
<tr>
<td><strong>Bold Arial</strong></td>
<td><strong>Keywords</strong> <em>(in the description of the syntax and in the text)</em></td>
</tr>
<tr>
<td></td>
<td>e.g. Rule ::= { ruleName : } { <strong>when</strong> antecedentCondition <strong>then</strong> }</td>
</tr>
<tr>
<td></td>
<td>consequentCondition</td>
</tr>
<tr>
<td></td>
<td>{ <strong>errorcode</strong> errorCode }</td>
</tr>
<tr>
<td></td>
<td>{ <strong>errorlevel</strong> errorLevel }</td>
</tr>
<tr>
<td></td>
<td>e.g. “....The <strong>rename</strong> operator allows to rename one or more Components...”</td>
</tr>
<tr>
<td><strong>Italics Arial</strong></td>
<td><strong>Optional Parameter</strong> <em>(in the description of the syntax)</em></td>
</tr>
<tr>
<td></td>
<td>e.g. <strong>substr</strong> ( op, start, length )</td>
</tr>
<tr>
<td><strong>Underlined Arial</strong></td>
<td><em>Sub-expressions</em></td>
</tr>
<tr>
<td><strong>Normal font Arial</strong></td>
<td><em>The operator’s syntax (excluded the keywords, the optional Parameters and the sub-expressions)</em></td>
</tr>
<tr>
<td></td>
<td>e.g. <strong>length</strong> ( &quot;Hello, World!&quot; )</td>
</tr>
<tr>
<td></td>
<td><em>The examples of invocation of the operators</em></td>
</tr>
<tr>
<td></td>
<td>e.g. <strong>length</strong> ( &quot;Hello, World!&quot; )</td>
</tr>
<tr>
<td></td>
<td><em>Optional and Mandatory Parameters</em> <em>(in the text)</em></td>
</tr>
</tbody>
</table>
|                  | e.g. “......If comp is a Measure in op, then in the result ......”
Abbreviations for the names of the artefacts

The names of the artefacts operated by the VTL-ML come from the VTL IM. In their turn, the names of the VTL IM artefacts are derived as much as possible from the names of the GSIM IM artefacts, as explained in the User Manual.

If the complete names are long, the VTL IM suggests also a compact name, which can be used in place of the complete name in case there is no ambiguity (for example, “Set” instead than “Value Domain Subset”, “Component” instead than “Data Set Component” and so on); moreover, to make the descriptions more compact, a number of abbreviations, usually composed of the initials (in capital case) or the first letters of the words of artefact names, are adopted in this manual:

<table>
<thead>
<tr>
<th>Complete name</th>
<th>Compact name</th>
<th>Abbreviation</th>
</tr>
</thead>
<tbody>
<tr>
<td>Data Set</td>
<td>Data Set</td>
<td>DS</td>
</tr>
<tr>
<td>Data Point</td>
<td>Data Point</td>
<td>DP</td>
</tr>
<tr>
<td>Identifier Component</td>
<td>Identifier</td>
<td>Id</td>
</tr>
<tr>
<td>Measure Component</td>
<td>Measure</td>
<td>Me</td>
</tr>
<tr>
<td>Attribute Component</td>
<td>Attribute</td>
<td>At</td>
</tr>
<tr>
<td>Data Set Component</td>
<td>Component</td>
<td>Comp</td>
</tr>
<tr>
<td>Value Domain Subset</td>
<td>Subset or Set</td>
<td>Set</td>
</tr>
<tr>
<td>Value Domain</td>
<td>Domain</td>
<td>VD</td>
</tr>
</tbody>
</table>

A positive integer suffix (with or without an underscore) can be added in the end to distinguish more than one instance of the same artefact (e.g., DS_1, DS_2, ..., DS_N, Me1, Me2, ...MeN). The suffix “r” stands for the result of a Transformation (e.g., DS_r).

Conventions for describing the operators’ syntax

Each VTL operator has an explanatory name, which recalls the operator function (e.g., “Greater than”) and a syntactical symbol, which is used to invoke the operator (e.g., “>”). The operator symbol may also be alphabetic, always lowercase (e.g., round).

In the VTL-DL, the operator symbol is the keyword define followed by the name of the object to be defined. The complete operator symbol is therefore a compound lowercase sentence (e.g., define operator).

In the VTL-ML, the operator symbol does not contain spaces and may be either a sequence of special characters (like +, -, >=, <= and so on) or a text keyword (e.g., and, or, not). The keyword may be compound with underscores as separators (e.g., exists_in).

Each operator has a syntax, which is a set of formal rules to invoke the operator correctly. In this document, the syntax of the operators is formally described by means of a meta-syntax which is not part of the VTL language, but has only presentation purposes.

The meta-syntax describes the syntax of the operators by means of meta-expressions, which define how the invocations of the operators must be written. The meta-expressions contain the symbol of the operator (e.g., “join”), the possible other keywords to denote special parameters (e.g., using), other symbols to be used (e.g., parentheses, commas), the named formal parameters (e.g., multiplicand and multiplier for the multiplication).

As for the typographic style, in order to distinguish between the syntax symbols (which are used in the operator invocations) and meta-syntax symbols (used just for explanatory purposes, and not actually used in invocations), the syntax symbols are in boldface (i.e., the operator symbol, the special keywords, the possible parenthesis, commas an so on). The names of the generic operands (e.g., multiplicand, multiplier) are in Roman type, even if they are part of the syntax.

The meta-expression can be very simple, for example the meta-expression for the addition is:

\[ \text{op1} + \text{op2} \]

This means that the addition has two operands (op1, op2) and is invoked by specifying the name of the first addendum (op1), then the addition symbol (+) followed by the name of the second addendum (op2).

In this example, all the three parts of the meta-expression are fixed. In other cases, the meta-expression can be more complex and made of optional, alternative or repeated parts.

In the simple cases, the optional parts are denoted by using the italic face, for example:
The expression above implies that in the `substr` operator the `start` and `length` operands are optional. In the invocation, a non-specified optional operand is substituted by an underscore or, if it is in the end of the invocation, can be omitted. Hence the following syntaxes are all formally correct:

```
substr ( op, start, length )
substr ( op, start )
substr ( op, _ , length )
substr ( op )
```

In more complex cases, a regular expression style is used to denote the parts (sub-expressions) of the meta-expression that are optional, alternative or repeated. In particular, braces denote a sub-expression; a vertical bar (or sometimes named “pipe”) within braces denotes possible alternatives; an optional trailing number, following the braces, specifies the number of possible repetitions.

- non-optional : non-optional sub-expression (text without braces)
- { optional } : optional sub-expression (zero or 1 occurrence)
- { non-optional } : non-optional sub-expression (just 1 occurrence)
- { one-or-more } : sub-expression repeatable from 1 to many occurrences
- { zero-or-more } : sub-expression repeatable from 0 to many occurrences
- { part1 | part2 | part3 } : optional alternative sub-expressions (zero or 1 occurrence)
- { part1 | part2 | part3 } : alternative sub-expressions (just 1 occurrence)
- { part1 | part2 | part3 }+ : alternative sub-expressions, from 1 to many occurrences
- { part1 | part2 | part3 }* : alternative sub-expressions, from 0 to many occurrences

Moreover, to improve the readability, some sub-expressions (the underlined ones) can be referenced by their names and separately defined, for example a meta-expression can take the following form:

```
sub-exp1-text sub-exp2-name ... sub-expN-name sub-expn-text
sub-exp2-name ::= sub-exp2-text
... possible others ...
sub-expN-name ::= sub-expn-text
```

In this representation of a meta-expression:

- The first line is the text of the meta-expression
- `sub-exp1-text, sub-expn-text` are sub-expressions directly written in the meta-expression
- `sub-exp2-name, ... sub-expN-name` are identifiers of sub-expressions.
- `sub-exp2-text, ... sub-expn-text` are subexpression written separately from the meta-expression.
- The symbol `::=` means “is defined as” and denotes the assignment of a sub-expression-text to a sub-expression-name.

The following example shows the definition of the syntax of the operators for removing the leading and/or the trailing whitespaces from a string:

```
Meta-expression ::= { trim | ltrim | rtrim } ( op )
```

The meta-expression above synthesizes that:

- `trim`, `ltrim`, `rtrim` are the operators’ symbols (reserved keywords);
- `()` are symbols of the operators syntax (reserved keywords);
- `op` is the only operand of the three operators;
- "{ }" and "[ ]" are symbols of the meta-syntax; in particular "[ ]" indicates that the three operators are alternative (a single invocation can contain only one of them) and "{ }" indicates that a single invocation contains just one of the shown alternatives;

From this template, it is possible to infer some valid possible invocations of the operators:

```
ltrim ( DS_2 )
rtrim ( DS_3 )
```

In these invocations, `ltrim` and `rtrim` are the symbols of the invoked operator and `DS_2` and `DS_3` are the names of the specific Data Sets which are operands respectively of the former and the latter invocation.
Description of the data types of operands and result

This section contains a brief legend of the meaning of the symbols used for describing the possible types of operands and results of the VTL operators. For a complete description of the VTL data types, see the chapter "VLT Data Types" in the User Manual.

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Meaning</th>
<th>Example</th>
<th>Example meaning</th>
</tr>
</thead>
<tbody>
<tr>
<td>parameter :: type2</td>
<td>parameter is of the type2</td>
<td>param1 :: string</td>
<td>param1 is of type string</td>
</tr>
<tr>
<td>type1</td>
<td>alternative types</td>
<td>dataset</td>
<td>component</td>
</tr>
<tr>
<td>type1</td>
<td>scalar type2 restricts type1</td>
<td>measure &lt; string&gt;</td>
<td>Measure of string type</td>
</tr>
<tr>
<td>type1 _ (underscore)</td>
<td>type1 can appear just once</td>
<td>measure &lt; string&gt; _</td>
<td>just one string Measure</td>
</tr>
<tr>
<td>type1 elementName</td>
<td>predetermined element of type1</td>
<td>measure &lt; string&gt; my_text</td>
<td>just one string Measure named &quot;my_text&quot;</td>
</tr>
<tr>
<td>type1 _ +</td>
<td>type1 can appear one or more times</td>
<td>measure &lt; string&gt; _+</td>
<td>one or more string Measures</td>
</tr>
<tr>
<td>type1 _ *</td>
<td>type1 can appear zero, one or more times</td>
<td>measure &lt; string&gt; _*</td>
<td>zero, one or more string Measures</td>
</tr>
<tr>
<td>dataset { type_constraint }</td>
<td>Type_constraint restricts the dataset type</td>
<td>dataset { measure &lt; string &gt; _+ }</td>
<td>Dataset having one or more string Measures</td>
</tr>
<tr>
<td>( t_1 \ast t_2 \ast \ldots \ast t_n )</td>
<td>Product of the types ( t_1, t_2, \ldots, t_n )</td>
<td>string \ast integer \ast boolean</td>
<td>triple of scalar values made of a string, an integer and a boolean value</td>
</tr>
<tr>
<td>( t_1 \rightarrow t_2 )</td>
<td>Operator from ( t_1 ) to ( t_2 )</td>
<td>string -&gt; number</td>
<td>Operator having input string and output number</td>
</tr>
<tr>
<td>ruleset { type_constraint }</td>
<td>Type_constraint restricts the ruleset type</td>
<td>hierarchical { geo_area }</td>
<td>hierarchical ruleset defined on geo_area</td>
</tr>
<tr>
<td>set &lt; t &gt;</td>
<td>Set of elements of type &quot;t&quot;</td>
<td>set &lt; dataset &gt;</td>
<td>set of datasets</td>
</tr>
</tbody>
</table>

Moreover, the word “name” in the data type description denotes the fact that the argument of the invocation can contain only the name of an artefact of such a type but not a sub-expression. For example:

```
comp :: name < component < string >>
```

Means that the argument passed for the input parameter \( \text{comp} \) can be only the name of a Component of the basic scalar type \( \text{string} \). The argument passed for \( \text{comp} \) cannot be a component expression.

The word “name” added as a suffix to the parameter name means the same (for example if the parameter above is called \( \text{comp}_\text{name} \)).
## VTL-ML Operators

<table>
<thead>
<tr>
<th>Name</th>
<th>Symbol</th>
<th>Syntax</th>
<th>Description</th>
<th>Notation</th>
<th>Input parameters type</th>
<th>Result type</th>
<th>Behaviour</th>
</tr>
</thead>
<tbody>
<tr>
<td>Parentheses</td>
<td>()</td>
<td>( op )</td>
<td>Override the default evaluation order of the operators</td>
<td>Func.</td>
<td>op :: dataset</td>
<td>component</td>
<td>scalar</td>
</tr>
<tr>
<td>Persistent assignment</td>
<td>&lt;-</td>
<td>re &lt;- op</td>
<td>Assigns an Expression to a persistent model artefact</td>
<td>Infix</td>
<td>re :: name</td>
<td>op :: dataset</td>
<td>empty</td>
</tr>
<tr>
<td>Non persistent assignment</td>
<td>:=</td>
<td>re := op</td>
<td>Assigns an Expression to a non persistent model artefact</td>
<td>Infix</td>
<td>re :: name</td>
<td>op :: dataset</td>
<td>scalar</td>
</tr>
<tr>
<td>Membership</td>
<td>#</td>
<td>ds#comp</td>
<td>Identifies a Component within a Data Set</td>
<td>Infix</td>
<td>ds :: dataset</td>
<td>comp :: name&lt;component&gt;</td>
<td>dataset</td>
</tr>
<tr>
<td>User defined operator call</td>
<td></td>
<td>operator_name {{ argument {, argument}^* }}</td>
<td>Invokes a user-defined operator passing the arguments</td>
<td>Func.</td>
<td>operatorName :: name</td>
<td>argument :: user-defined operator parameters data type</td>
<td>user-defined result data type</td>
</tr>
<tr>
<td>Evaluation of an external routine</td>
<td>eval</td>
<td>eval ( externalRoutineName { (argument) {, argument}^* }, language, returns outputType )</td>
<td>Evaluates an external routine</td>
<td>Func.</td>
<td>externalRoutineName :: string</td>
<td>argument :: any expression</td>
<td>language :: string</td>
</tr>
<tr>
<td>Type conversion</td>
<td>cast</td>
<td><code>cast (op . scalarType (. mask ))</code></td>
<td>converts to the specified data type</td>
<td>Func.</td>
<td><code>op :: dataset (measure&lt;scalar&gt;_)</code></td>
<td>On two scalars, DSs or DSCs</td>
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</tr>
<tr>
<td>Join</td>
<td><code>inner_joi n, left_join, full_join, cross_joi n</code></td>
<td><code>joinOperator (ds { as alias }, ds { as alias })</code>*</td>
<td>ds :: dataset</td>
<td>ds :: dataset</td>
<td></td>
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<td></td>
<td><code>alias :: name</code></td>
<td>alias :: name</td>
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<td><code>usingId :: name &lt; component&gt;</code></td>
<td>usingId :: name &lt; component&gt;</td>
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<td><code>filterCondition :: component&lt;boolean&gt;</code></td>
<td>filterCondition :: component&lt;boolean&gt;</td>
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<td><code>applyExpr :: dataset</code></td>
<td>applyExpr :: dataset</td>
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<td><code>calcComp :: name&lt;component&gt;</code></td>
<td>calcComp :: name&lt;component&gt;</td>
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<td><code>calcExpr :: component&lt;scalar&gt;</code></td>
<td>calcExpr :: component&lt;scalar&gt;</td>
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<td><code>aggrComp :: name&lt;component&gt;</code></td>
<td>aggrComp :: name&lt;component&gt;</td>
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<td><code>aggrExpr :: component&lt;scalar&gt;</code></td>
<td>aggrExpr :: component&lt;scalar&gt;</td>
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<td><code>conversionId :: name &lt; identifier&gt;</code></td>
<td>conversionId :: name &lt; identifier&gt;</td>
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<td><code>havingCondition :: component&lt;boolean&gt;</code></td>
<td>havingCondition :: component&lt;boolean&gt;</td>
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<td></td>
<td><code>comp :: name &lt; component&gt;</code></td>
<td>comp :: name &lt; component&gt;</td>
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<td><code>compFrom :: component&lt;scalar&gt;</code></td>
<td>compFrom :: component&lt;scalar&gt;</td>
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<td><code>compTo :: component&lt;scalar&gt;</code></td>
<td>compTo :: component&lt;scalar&gt;</td>
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<td></td>
</tr>
<tr>
<td>String concatenation</td>
<td>`</td>
<td></td>
<td>`</td>
<td>`op1</td>
<td></td>
<td>op2`</td>
<td>Concatenates two strings</td>
</tr>
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<td></td>
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<td></td>
<td></td>
<td></td>
<td>`</td>
<td>component&lt;string&gt;`</td>
<td></td>
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<td>`</td>
<td>string`</td>
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<td></td>
<td>`</td>
<td>string`</td>
<td></td>
</tr>
</tbody>
</table>

**Type conversion**
- **cast**
  - `cast (op . scalarType (. mask ))` converts to the specified data type
  - Func.
    - `op :: dataset (measure<scalar>_)`
    - `| component<scalar>`
    - `| scalar`
    - `scalarType :: scalar type`
    - `mask :: string`
<table>
<thead>
<tr>
<th>Function</th>
<th>Syntax</th>
<th>Description</th>
<th>Arguments</th>
<th>Output</th>
<th>Examples</th>
</tr>
</thead>
<tbody>
<tr>
<td>Whitespace removal</td>
<td><code>trim</code>, <code>ltrim</code>, <code>rtrim</code></td>
<td>Removes trailing or/and leading whitespace from a string</td>
<td><code>op</code></td>
<td>`dataset { measure&lt;string&gt;</td>
<td>component&lt;string&gt;</td>
</tr>
<tr>
<td>Character case conversion</td>
<td><code>upper</code>, <code>lower</code></td>
<td>Converts the character case of a string in upper or lower case</td>
<td><code>op</code></td>
<td>`dataset { measure&lt;string&gt;</td>
<td>component&lt;string&gt;</td>
</tr>
<tr>
<td>Sub-string extraction</td>
<td><code>substr</code></td>
<td>Extracts the substring that starts in a specified position and has a specified length</td>
<td><code>op</code>, <code>start</code>, <code>length</code></td>
<td>`dataset { measure&lt;string&gt;</td>
<td>component&lt;string&gt;</td>
</tr>
<tr>
<td>String pattern replacement</td>
<td><code>replace</code></td>
<td>Replaces a specified string-pattern with another one</td>
<td><code>op</code>, <code>pattern1</code>, <code>pattern2</code></td>
<td>`dataset { measure&lt;string&gt;</td>
<td>component&lt;string&gt;</td>
</tr>
</tbody>
</table>
| String pattern location | instr | instr(op, pattern, start, occurrence) | Returns the location of a specified string-pattern | Func. | op ::
dataset { measure<string> _+ } | component<string>
| pattern :: component<string>
| string
| start :: component<int> >= 1>
| integer[>= 1]
| occurrence :: component<int[>= 1]>
| integer[>= 1]
| dataset
{measure<integer[>=0]>
| component<integer[>= 0]>
| integer[>= 0]}
| Changing data type

| String length | length | length(op) | Returns the length of a string | Func. | op ::
dataset { measure<string> _- } | component<string>
| string
| dataset
{measure<integer[>=0]>
| component<integer[>= 0]>
| integer[>= 0]}
| Changing data type

| Unary plus | + | + op | Replicates the operand with the sign unaltered | Infix | op ::
dataset { measure<number> _+ } | component<number>
| number
| dataset
{measure<number> _+ }
| component<number>
| number
| On one scalar, DS or DSC

| Unary minus | - | - op | Replicates the operand with the sign changed | Infix | op ::
dataset { measure<number> _- } | component<number>
| number
| dataset
{measure<number> _- }
| component<number>
| number
| On one scalar, DS or DSC

| Addition | + | op1 + op2 | Sums two numbers | Infix | op1, op2::
dataset { measure<number> _+ } | component<number>
| number
| dataset
{measure<number> _+ }
| component<number>
| number
| On two scalars, DSs or DSCs

| Subtraction | - | op1 - op2 | Subtracts two numbers | Infix | op1, op2::
dataset { measure<number> _- } | component<number>
| number
| dataset
{measure<number> _- }
| component<number>
| number
| On two scalars, DSs or DSCs

| Multiplication | * | op1 * op2 | Multiplies two numbers | Infix | op1, op2::
dataset { measure<number> _* } | component<number>
| number
| dataset
{measure<number> _* }
| component<number>
| number
| On two scalars, DSs or DSCs

| Division | / | op1 / op2 | Divides two numbers | Infix | op1, op2::
dataset { measure<number> _/ } | component<number>
| number
| dataset
{measure<number> _/ }
| component<number>
| number
| On two scalars, DSs or DSCs

Changing data type
<table>
<thead>
<tr>
<th>Operation</th>
<th>Symbol</th>
<th>Description</th>
<th>Function Signature</th>
<th>Dataset Signature</th>
<th>Notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>Modulo</td>
<td>mod</td>
<td>Calculates the remainder of a number divided by a certain divisor</td>
<td><code>mod(op1, op2)</code></td>
<td><code>{ measure&lt;number&gt; _+ } \mid component&lt;number&gt; \mid number</code></td>
<td>On two scalar, DS or DSC</td>
</tr>
<tr>
<td>Rounding</td>
<td>round</td>
<td>Rounds a number to a certain digit</td>
<td><code>round(op, numDigit)</code></td>
<td><code>{ measure&lt;number&gt; _+ } \mid component&lt;number&gt; \mid number</code></td>
<td>On one DS or on two scalars or DSC</td>
</tr>
<tr>
<td>Truncation</td>
<td>trunc</td>
<td>Truncates a number to a certain digit</td>
<td><code>trunc(op, numDigit)</code></td>
<td><code>{ measure&lt;number&gt; _+ } \mid component&lt;number&gt; \mid number</code></td>
<td>On one DS or on two scalars or DSC</td>
</tr>
<tr>
<td>Ceiling</td>
<td>ceil</td>
<td>Returns the smallest integer which is greater or equal than a number</td>
<td><code>ceil(op)</code></td>
<td><code>{ measure&lt;integer&gt; _+ } \mid component&lt;integer&gt; \mid integer</code></td>
<td>On one scalar, DS or DSC</td>
</tr>
<tr>
<td>Floor</td>
<td>floor</td>
<td>Returns the greater integer which is smaller or equal than a number</td>
<td><code>floor(op)</code></td>
<td><code>{ measure&lt;integer&gt; _+ } \mid component&lt;integer&gt; \mid integer</code></td>
<td>On one scalar, DS or DSC</td>
</tr>
<tr>
<td>Absolute value</td>
<td>abs</td>
<td>Calculates the absolute value of a number</td>
<td><code>abs(op)</code></td>
<td><code>{ measure&lt;number&gt; _+ } \mid component&lt;number&gt; \mid number</code></td>
<td>On one scalar, DS or DSC</td>
</tr>
<tr>
<td>Exponential</td>
<td>exp</td>
<td>Raises e (base of the natural logarithm) to a number</td>
<td><code>exp(op)</code></td>
<td><code>{ measure&lt;number&gt; _+ } \mid component&lt;number&gt; \mid number</code></td>
<td>On one scalar, DS or DSC</td>
</tr>
<tr>
<td>Natural logarithm</td>
<td>ln</td>
<td>( \ln(\text{op}) )</td>
<td>Calculates the natural logarithm of a number</td>
<td>Func.</td>
<td>op :: dataset (measure&lt;number&gt;[&gt;0]&gt;_+ )</td>
</tr>
<tr>
<td>---</td>
<td>---</td>
<td>---</td>
<td>---</td>
<td>---</td>
<td>---</td>
</tr>
<tr>
<td>Power</td>
<td>power</td>
<td>( \text{power}(\text{base}, \text{exponent}) )</td>
<td>Raises a number to a certain exponent</td>
<td>Func.</td>
<td>base :: dataset (measure&lt;number&gt;[&gt;1]&gt;_+ )</td>
</tr>
<tr>
<td>Logarithm</td>
<td>log</td>
<td>( \log(\text{op}, \text{num}) )</td>
<td>Calculates the logarithm of a number to a certain base</td>
<td>Func.</td>
<td>op :: dataset (measure&lt;number&gt;[&gt;1]&gt;_+ )</td>
</tr>
<tr>
<td>Square root</td>
<td>sqrt</td>
<td>( \sqrt{\text{op}} )</td>
<td>Calculates the square root of a number</td>
<td>Func.</td>
<td>op :: dataset (measure&lt;number&gt;[&gt;=0]&gt;_+ )</td>
</tr>
<tr>
<td>Equal to</td>
<td>( = )</td>
<td>left = right</td>
<td>Verifies if two values are equal</td>
<td>Infix</td>
<td>left, right :: dataset (measure&lt;scalar&gt;_)</td>
</tr>
<tr>
<td>Not equal to</td>
<td>( &lt;&gt; )</td>
<td>left &lt;&gt; right</td>
<td>Verifies if two values are not equal</td>
<td>Infix</td>
<td>left, right :: dataset (measure&lt;scalar&gt;_)</td>
</tr>
<tr>
<td>Greater than</td>
<td>( &gt; )</td>
<td>left ( ( &gt; \mid &gt;= )) right</td>
<td>Verifies if a first value is greater (or equal) than a second value</td>
<td>Infix</td>
<td>left, right :: dataset (measure&lt;scalar&gt;_)</td>
</tr>
<tr>
<td>Less than</td>
<td>( &lt; )</td>
<td>left ( ( &lt; \mid &lt;= )) right</td>
<td>Verifies if a first value is less (or equal) than a second value</td>
<td>Infix</td>
<td>left, right :: dataset (measure&lt;scalar&gt;_)</td>
</tr>
<tr>
<td>Between</td>
<td>between</td>
<td>between( op, from, to )</td>
<td>Verify if a value belongs to a range of values</td>
<td>Func. op :: dataset {measure&lt;scalar&gt;._}</td>
<td>component&lt;scalar&gt;</td>
</tr>
<tr>
<td>-----------</td>
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<td>------------------------</td>
<td>------------------------------------------------</td>
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<td>-----------------</td>
</tr>
<tr>
<td>Element of</td>
<td>in</td>
<td>op in collection</td>
<td>Verifies if a value belongs to a set of values</td>
<td>Infix op :: dataset {measure&lt;scalar&gt;._}</td>
<td>component&lt;scalar&gt;</td>
</tr>
<tr>
<td></td>
<td></td>
<td>op not_in collection</td>
<td>Verifies if a value does not belong to a set of values</td>
<td>Infix op :: dataset {measure&lt;scalar&gt;._}</td>
<td>component&lt;scalar&gt;</td>
</tr>
<tr>
<td>Match_characters</td>
<td>match_characters</td>
<td>match_characters(op, pattern)</td>
<td>Verifies if a value respects or not a pattern</td>
<td>Func. op :: dataset {measure&lt;string&gt;._}</td>
<td>component&lt;string&gt;</td>
</tr>
<tr>
<td>Isnull</td>
<td>isnull</td>
<td>isnull(op)</td>
<td>Verifies if a value is NULL</td>
<td>Func. op :: dataset {measure&lt;scalar&gt;._}</td>
<td>component&lt;scalar&gt;</td>
</tr>
<tr>
<td>Exists_in</td>
<td>exists_in</td>
<td>exists_in(op1, op2, retain)</td>
<td>As for the common identifiers of op1 and op2, verifies if the combinations of values of op1 exist in op2.</td>
<td>Func. op1,op2 :: dataset</td>
<td>dataset {measure&lt;boolean&gt; bool_var}</td>
</tr>
<tr>
<td>Logical conjunction</td>
<td>and</td>
<td>op1 and op2</td>
<td>Calculates the logical AND</td>
<td>op1,op2 :: dataset {measure&lt;boolean&gt;._}</td>
<td>component&lt;boolean&gt;</td>
</tr>
<tr>
<td>Logical disjunction</td>
<td>or</td>
<td>op1 or op2</td>
<td>Calculates the logical OR</td>
<td>op1,op2 :: dataset {measure&lt;boolean&gt;._}</td>
<td>component&lt;boolean&gt;</td>
</tr>
<tr>
<td><strong>Exclusive disjunction</strong></td>
<td>xor</td>
<td>op1 xor op2</td>
<td>Calculates the logical XOR</td>
<td>op1 xor op2</td>
<td>dataset (measure:boolean)</td>
</tr>
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<td>--------------------------</td>
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</tr>
<tr>
<td><strong>Logical negation</strong></td>
<td>not</td>
<td>not op</td>
<td>Calculates the logical NOT</td>
<td>op</td>
<td>dataset (measure:boolean)</td>
</tr>
<tr>
<td><strong>Period indicator</strong></td>
<td>period_indicator</td>
<td>period_indicator (op)</td>
<td>extracts the period indicator from a time_period value</td>
<td>Func. op</td>
<td>dataset (measure&lt;time_period&gt;</td>
</tr>
<tr>
<td><strong>Fill time series</strong></td>
<td>fill_time_series</td>
<td>fill_time_series (op, limitsMethod)</td>
<td>limitsMethod ::= single</td>
<td>all</td>
<td>Replaces each missing data point in the input Data Set</td>
</tr>
<tr>
<td><strong>Flow to stock</strong></td>
<td>flow_to_stock</td>
<td>flow_to_stock (op)</td>
<td>Transforms from a flow interpretation of a Data Set to stock</td>
<td>Func. op</td>
<td>dataset (identifier&lt;time&gt;</td>
</tr>
<tr>
<td><strong>Stock to flow</strong></td>
<td>stock_to_flow</td>
<td>stock_to_flow (op)</td>
<td>Transforms from stock to flow interpretation of a Data Set</td>
<td>Func. op</td>
<td>dataset (identifier&lt;time&gt;</td>
</tr>
<tr>
<td><strong>Time shift</strong></td>
<td>timeshift</td>
<td>timeshift (op, shiftNumber)</td>
<td>Shifts the time component of a specified range of time</td>
<td>Func. op</td>
<td>dataset (identifier&lt;time&gt;</td>
</tr>
<tr>
<td><strong>Time aggregation</strong></td>
<td>time_agg</td>
<td>time_agg (periodIndTo, periodIndFrom)</td>
<td>converts the time values from higher to lower frequency values</td>
<td>Func. op</td>
<td>dataset (identifier&lt;time&gt;</td>
</tr>
<tr>
<td>Actual time</td>
<td>current_date</td>
<td>current_date ()</td>
<td>returns the current date</td>
<td>Func.</td>
<td>date</td>
</tr>
<tr>
<td>-------------</td>
<td>--------------</td>
<td>-----------------</td>
<td>-------------------------</td>
<td>-------</td>
<td>------</td>
</tr>
<tr>
<td>Union</td>
<td>union</td>
<td>union (dsList)</td>
<td>Computes the union of N datasets</td>
<td>Func. ds :: dataset</td>
<td>dataset</td>
</tr>
<tr>
<td>Intersection</td>
<td>intersect</td>
<td>intersect (dsList)</td>
<td>Computes the intersection of N datasets</td>
<td>Func. ds :: dataset</td>
<td>dataset</td>
</tr>
<tr>
<td>Set difference</td>
<td>setdiff</td>
<td>setdiff (ds1, ds2)</td>
<td>Computes the differences of two datasets</td>
<td>Func. ds1, ds2 :: dataset</td>
<td>dataset</td>
</tr>
<tr>
<td>Symmetric difference</td>
<td>symdiff</td>
<td>symdiff (ds1, ds2)</td>
<td>Computes the symmetric difference of two datasets</td>
<td>Func. ds1, ds2 :: dataset</td>
<td>dataset</td>
</tr>
</tbody>
</table>

### Hierarchical roll-up

**hierarchy** (op, hr {condition condComp (, condComp)*} {rule ruleComp (, mode (input) (output))})

- **condComp** ::= component (, component)*
- **mode** ::= non_null | non_zero | partial_null | partial_zero | always_null | always_zero
- **input** ::= dataset | rule | rule_priority
- **output** ::= computed | all

Aggregates data using a hierarchical ruleset

**Func.**

```
op := dataset(measure<number> _)
hr := name < hierarchical >
condComp := name < component >
ruleComp := name < identifier >
dataset(measure<number> _)
```

**Specific**

### Aggregate invocation

**aggregateOperator**

- **aggregateOperator** (firstOperand (, additionalOperand)* (groupingClause) )

In a Component expression within an aggr clause

- **aggregateOperator** (firstOperand (, additionalOperand)* (groupingClause))

- **aggregateOperator** := avg | count | max | median | min | stddev_pop | stddev_samp | sum | var_pop | var_samp

**groupingClause** ::= (group by groupingId (, groupingId)* | group except groupingId (, groupingId)* | group all conversionExpr)* |

- (having havingCondition)

Set of statistical functions used to aggregate data

**Func.**

```
firstOperand := dataset | component
additionalOperand := type of the (possible) additional parameter of the aggregate Operator
groupingId := name < identifier >
conversionExpr := identifier
havingCondition := component<boolean>
dataset | component
```

**Specific**
### Analytic Invocation

Analytic invocation

| analyticOperator | analyticOperator ::= avg | count | max | median | min | stddev_pop | stddev_samp | sum | var_pop | var_samp | first_value | lead | rank | ratio_to_report |
|------------------|--------------------------------------------------|

**analyticOperator** ::= avg | count | max | median | min | stddev_pop | stddev_samp | sum | var_pop | var_samp | first_value | lead | rank | ratio_to_report

**analyticClause** ::= ( partitionClause ) ( orderClause ) ( windowClause )

**partitionClause** ::= partition by identifier { , identifier }*

**orderClause** ::= order by component { asc | desc } { , component { asc | desc } }*

**windowClause** ::= ( data points | range ) 1 between limitClause and limitClause

**limitClause** ::= ( num preceding | num following | current data point | unbounded preceding | unbounded following )

---

### Check Datapoint

Check datapoint

<table>
<thead>
<tr>
<th>check_datapoint</th>
<th>check_datapoint ( op , dpr { components listComp } ( output output ) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>listComp ::= comp ( , comp )*</td>
<td></td>
</tr>
<tr>
<td>output ::= invalid</td>
<td>all</td>
</tr>
</tbody>
</table>

Applies one datapoint ruleset on a Data Set

Func. op :: dataset |

dpr :: name < datapoint > |

comp :: name < component >

data_set component

---

### Check Hierarchy

Check hierarchy

<table>
<thead>
<tr>
<th>check_hierarchy</th>
<th>check_hierarchy ( op , hr { condition condComp ( , condComp )* }</th>
</tr>
</thead>
<tbody>
<tr>
<td>condComp ::= non_null</td>
<td>non_zero</td>
</tr>
<tr>
<td>input ::= dataset</td>
<td>dataset_priority</td>
</tr>
<tr>
<td>output ::= invalid</td>
<td>all</td>
</tr>
</tbody>
</table>

Applies a hierarchical ruleset to a Data Set

Func. op :: dataset |

hr :: name < hierarchical > |
condComp :: name < component > |
ruleComp :: name < identifier > |

data_set |

---

### Check

Check

<table>
<thead>
<tr>
<th>check</th>
<th>check ( op { errorcode errorcode } ( errorlevel errorlevel )</th>
</tr>
</thead>
<tbody>
<tr>
<td>output ::= invalid</td>
<td>all</td>
</tr>
</tbody>
</table>

Checks if an expression verifies a condition

Func. op :: dataset |

errorcode :: errorcode_vd |
errorlevel :: errorlevel_vd |
imbalance :: number |

data_set |
<table>
<thead>
<tr>
<th>If then else</th>
<th></th>
<th></th>
<th>Makes alternative calculations according to a condition</th>
<th>Func.</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>if <em>...then else...</em></td>
<td>if condition <em>then</em> thenOperand <em>else</em> elseOperand</td>
<td></td>
<td></td>
<td>condition :: dataset { measure</td>
<td>boolean }</td>
<td>component</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>thenOperand :: dataset</td>
<td>component</td>
<td>scalar</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>elseOperand :: dataset</td>
<td>component</td>
<td>scalar</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Nvl</th>
<th>nvl</th>
<th>nvl (op1, op2)</th>
<th>Replaces the null value with a value.</th>
<th>Func.</th>
<th>op1, op2:: dataset</th>
<th>component</th>
<th>scalar</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>dataset</td>
<td>component</td>
<td>scalar</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Filtering Data Points</th>
<th>filter</th>
<th>op [filter condition]</th>
<th>Filter data using a Boolean condition</th>
<th>Clause</th>
<th>op :: dataset</th>
<th>filterCondition :: component</th>
<th>boolean</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>dataset</td>
<td></td>
<td>Specific</td>
</tr>
</tbody>
</table>

| Calculation of a Component | calc | op [calc (calcRole) calcComp := calcExpr | (calcRole) calcComp := calcExpr]*] | Calculates the values of a Structure Component | Clause | op :: dataset | calcComp :: name | component |
| --- | --- | --- | --- | --- | --- | --- | --- |
|  |  |  |  |  | calcExp :: component | scalar |

<table>
<thead>
<tr>
<th>Aggregation</th>
<th>aggr</th>
<th>op [aggr aggrClause (groupingClause)]</th>
<th>Aggregates using an aggregate operator</th>
<th>Clause</th>
<th>aggrComp :: name</th>
<th>component</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>aggrExp :: component</td>
<td>scalar</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>groupingId :: name</td>
<td>identifier</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>conversionExpr :: identifier</td>
<td>scalar</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>havingCondition :: component</td>
<td>boolean</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Maintaining Components</th>
<th>keep</th>
<th>op [keep comp (comp)*]</th>
<th>Keep list of components</th>
<th>Clause</th>
<th>op :: dataset</th>
<th>comp :: name</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>dataset</td>
<td>component</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Removal of Components</th>
<th>drop</th>
<th>op [drop comp (comp)*]</th>
<th>Drop list of components</th>
<th>Clause</th>
<th>op :: dataset</th>
<th>comp :: name</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>dataset</td>
<td>component</td>
</tr>
<tr>
<td>Change of Component name</td>
<td>rename</td>
<td>op [ rename comp_from to comp_to (,comp_from to comp_to)* ]</td>
<td>Rename components</td>
<td>Clause</td>
<td>op :: dataset</td>
<td>dataset</td>
</tr>
<tr>
<td>-------------------------</td>
<td>--------</td>
<td>-------------------------------------------------------------</td>
<td>------------------</td>
<td>--------</td>
<td>---------------</td>
<td>---------</td>
</tr>
<tr>
<td></td>
<td></td>
<td>comp_from :: name&lt;component&gt; comp_to :: name&lt;component&gt;</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Pivoting</td>
<td>pivot</td>
<td>op [ pivot identifier , measure ]</td>
<td>Transform identifier values to measures</td>
<td>Clause</td>
<td>op :: dataset</td>
<td>dataset</td>
</tr>
<tr>
<td></td>
<td></td>
<td>identifier :: name&lt;identifier&gt; measure :: name&lt;measure&gt;</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unpivoting</td>
<td>unpivot</td>
<td>op [ unpivot identifier , measure ]</td>
<td>Transform measures to identifier values</td>
<td>Clause</td>
<td>op :: dataset</td>
<td>dataset</td>
</tr>
<tr>
<td></td>
<td></td>
<td>identifier :: name&lt;identifier&gt; measure :: name&lt;measure&gt;</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Subspace</td>
<td>sub</td>
<td>op [ sub identifier = value (, identifier = value)* ]</td>
<td>Remove the specified identifiers by fixing a value for them</td>
<td>Clause</td>
<td>op :: dataset</td>
<td>dataset</td>
</tr>
<tr>
<td></td>
<td></td>
<td>identifier :: name&lt;identifier&gt; value :: scalar</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Within a single expression of the manipulation language, the operators are applied in sequence, according to the precedence order. Operators with the same precedence level are applied according to the default associativity rule. Precedence and associativity orders are reported in the following table.

<table>
<thead>
<tr>
<th>Evaluation order</th>
<th>Operator</th>
<th>Description</th>
<th>Default associativity rule</th>
</tr>
</thead>
<tbody>
<tr>
<td>I</td>
<td>()</td>
<td>Parentheses. To alter the default order.</td>
<td>None</td>
</tr>
<tr>
<td>II</td>
<td>VTL operators with functional syntax</td>
<td>VTL operators with functional syntax</td>
<td>Left-to-right</td>
</tr>
<tr>
<td>III</td>
<td>Clause Membership</td>
<td>Clause Membership</td>
<td>Left-to-right</td>
</tr>
<tr>
<td>IV</td>
<td>Unary plus, unary minus, not</td>
<td>Unary minus, Unary plus, Logical negation</td>
<td>None</td>
</tr>
<tr>
<td>V</td>
<td>* /</td>
<td>Multiplication, Division</td>
<td>Left-to-right</td>
</tr>
<tr>
<td>VI</td>
<td>+ -</td>
<td></td>
<td></td>
</tr>
<tr>
<td>VII</td>
<td>&gt; &gt;= &lt; &lt;= = &lt;&gt; in not in</td>
<td>Greater than, Less than, Equal-to, Not-equal-to, In a value list, Not in a value list</td>
<td>Left-to-right</td>
</tr>
<tr>
<td>VIII</td>
<td>and</td>
<td>Logical AND</td>
<td>Left-to-right</td>
</tr>
<tr>
<td>IX</td>
<td>or xor</td>
<td>Logical OR, Logical XOR</td>
<td>Left-to-right</td>
</tr>
<tr>
<td>X</td>
<td>if-then-else</td>
<td>Conditional (if-then-else)</td>
<td>None</td>
</tr>
</tbody>
</table>

Description of VTL Operators

The structure used for the description of the VTL-DL Operators is made of the following parts:

- **Operator name**, which is also used to invoke the operator
- **Semantics**: a brief description of the purpose of the operator
- **Syntax**: the syntax of the Operator (this part follows the conventions described in the previous section “Conventions for describing the operators’ syntax”)
- **Syntax description**: detailed explanation of the meaning of the various parts of the syntax
- **Parameters**: list of the input parameters and their types
• **Constraints**: additional constraints that are not specified with the meta-syntax and need a textual explanation

• **Semantic specifications**: detailed description of the semantics of the operator

• **Examples**: examples of invocation of the operator

The structure used for the description of the VTL-ML Operators is made of the following parts:

• **Operator name**, followed by the **operator symbol** (keyword) which is used to invoke the operator

• **Syntax**: the syntax of the Operator (this part follows the conventions described in the previous section “Conventions for describing the operators’ syntax”)

• **Input parameters**: list of all input parameters and the subexpressions with their meaning and the indication if they are mandatory or optional

• **Examples of valid syntaxes**: examples of syntactically valid invocations of the Operator

• **Semantics for scalar operations**: the behaviour of the Operator on scalar operands, which is the basic behaviour of the Operator

• **Input parameters type**: the formal description of the type of the input parameters (this part follows the conventions described in the previous section “Description of the data types of operands and results”)

• **Result type**: the formal description of the type of the result (this part follows the conventions described in the previous section “Description of the data types of operands and results”)

• **Additional constraints**: additional constraints that are not specified with the meta-syntax and need a textual explanation, including both possible semantic constraints under which the operation is possible or impossible, and syntactical constraint for the invocation of the Operator

• **Behaviour**: description of the behaviour of the Operator for non-scalar operations (for example operations at Data Set or at Component level). When the Operator belongs to a class of Operators having a common behaviour, the common behavior is described once for all in a section of the chapter “Typical behaviours of the ML Operators” and therefore this part describes only the specific aspect of the behaviour and contains a reference to the section where the common part of the behaviour is described.

• **Examples**: a series of examples of invocation and application of the operator in case of operations at Data Sets or at Component level.
VTL-DL - Rulesets

define datapoint ruleset

Semantics

The Data Point RuleSet contains Rules to be applied to each individual Data Point of a Data Set for validation purposes. These rulesets are also called “horizontal” taking into account the tabular representation of a Data Set (considered as a mathematical function), in which each (vertical) column represents a variable and each (horizontal) row represents a Data Point: these rulesets are applied on individual Data Points (rows), i.e., horizontally on the tabular representation.

Syntax

define datapoint ruleset rulesetName ( dpRulesetSignature ) is
dpRule *
end datapoint ruleset

dpRulesetSignature ::= valuedomain listValueDomains | variable listVariables
listValueDomains ::= valueDomain { as vdAlias } { , valueDomain { as vdAlias } }
listVariables ::= variable { as varAlias } { , variable { as varAlias } }
dpRule ::= { ruleName : } { when antecedentCondition then } consequentCondition

Syntax description

rulesetName the name of the Data Point RuleSet to be defined.
dpRulesetSignature the Cartesian space of the RuleSet (signature of the RuleSet), which specifies either the Value Domains or the Represented Variables (see the information model) on which the RuleSet is defined. If valuedomain is specified then the RuleSet is applicable to Data Sets having Components that take values on the specified Value Domains. If variable is specified then the RuleSet is applicable to Data Sets having the specified Variables as Components.

valueDomain a Value Domain on which the RuleSet is defined.
vdAlias an (optional) alias assigned to a Value Domain and valid only within the RuleSet, this can be used for the sake of compactness in writing the Rules. If an alias is not specified then the name of the Value Domain (parameter valueDomain) is used in the body of the rules.

variable an (optional) alias assigned to a Variable and valid only within the RuleSet, this can be used for the sake of compactness in writing the Rules. If an alias is not specified then the name of the Variable (parameter valueDomain) is used in the body of the Rules.

dpRule a Data Point Rule, as defined in the following parameters.

ruleName the name assigned to the specific Rule within the RuleSet. If the RuleSet is used for validation then the ruleName identifies the validation results of the various Rules of the RuleSet. The ruleName is optional and, if not specified, is assumed to be the progressive order number of the Rule in the RuleSet. However please note that, if ruleName is omitted, then the Rule names can change in case the RuleSet is modified, e.g., if new Rules are added or existing Rules are deleted, and therefore the users that interpret the validation results must be aware of these changes.

antecedentCondition a boolean expression to be evaluated for each single Data Point of the input Data Set. It can contain Values of the Value Domains or Variables specified in the RuleSet signature and constants; all the VTL-ML component level operators are allowed. If omitted then antecedentCondition is assumed to be TRUE.

consequentCondition a boolean expression to be evaluated for each single Data Point of the input Data Set when the antecedentCondition evaluates to TRUE (as mentioned, missing antecedent
conditions are assumed to be TRUE). It contains Values of the Value Domains or Variables specified in the Ruleset signature and constants; all the VTL-ML component level operators are allowed. A consequent condition equal to FALSE is considered as a non-valid result.

- **errorCode**
  a literal denoting the error code associated to the rule, to be assigned to the possible non-valid results in case the Rule is used for validation. If omitted then no error code is assigned (NULL value). VTL assumes that a Value Domain errorcode_vd of error codes exists in the Information Model and contains all possible error codes: the errorCode literal must be one of the possible Values of such a Value Domain. VTL assumes also that a Variable errorCode for describing the error codes exists in the IM and is a dependent variable of the Data Sets which contain the results of the validation.

- **errorLevel**
  a literal denoting the error level (severity) associated to the rule, to be assigned to the possible non-valid results in case the Rule is used for validation. If omitted then no error level is assigned (NULL value). VTL assumes that a Value Domain errorlevel_vd of error levels exists in the Information Model and contains all possible error levels: the errorLevel literal must be one of the possible Values of such a Value Domain. VTL assumes also that a Variable errorlevel for describing the error levels exists in the IM and is a dependent variable of the Data Sets which contain the results of the validation.

### Parameters

- **rulesetName** :: name <ruleset >
- **valueDomain** :: name < valuedomain >
- **vdAlias** :: name
- **variable** :: name
- **varAlias** :: name
- **ruleName** :: name
- **antecedentCondition** :: boolean
- **consequentCondition** :: boolean
- **errorCode** :: errorcode_vd
- **errorLevel** :: errorlevel_vd

### Constraints

- antecedentCondition and consequentCondition can refer only to the Value Domains or Variables specified in the dpRulesetSignature.
- Either ruleName is specified for all the Rules of the Ruleset or for none.
- If specified, then ruleName must be unique within the Ruleset.

### Semantic specification

This operator defines a persistent Data Point Ruleset named rulesetName that can be used for validation purposes.

A Data Point Ruleset is a persistent object that contains Rules to be applied to the Data Points of a Data Set. The Data Point Rulesets can be invoked by the check_datapoint operator. The Rules are aimed at checking the combinations of values of the Data Set Components, assessing if these values fulfill the logical conditions expressed by the Rules themselves. The Rules are evaluated independently for each Data Point, returning a Boolean scalar value (i.e., TRUE for valid results and FALSE for non-valid results).

Each Rule contains an (optional) antecedentCondition boolean expression followed by a consequentCondition boolean expression and expresses a logical implication. Each Rule states that when the antecedentCondition evaluates to TRUE for a given Data Point, then the consequentCondition is expected to be TRUE as well. If this implication is fulfilled, the result is considered as valid (TRUE), otherwise as non-valid (FALSE). On the other side, if the antecedentCondition evaluates to FALSE, the consequentCondition does not applies and is not evaluated at all, and the result is considered as valid (TRUE). In case the antecedentCondition is absent then it is assumed to be always TRUE, therefore the consequentCondition is expected to evaluate to TRUE for all the Data Points. See an example below:

---

1 In order to apply the Ruleset to more Data Sets, these Data Sets must be composed together using the appropriate VTL operators in order to obtain a single Data Set.
The definition of a Ruleset comprises a **signature** (dpRulesetSignature), which specifies the Value Domains or Variables on which the Ruleset is defined and a set of Rules, that are the Boolean expressions to be applied to each Data Point. The **antecedentCondition** and **consequentCondition** of the Rules can refer only to the Value Domains or Variables of the Ruleset signature.

The Value Domains or the Variables of the Ruleset signature identify the space in which the rules are defined while each Rule provides for a criterion that demarcates the Set of valid combinations of Values inside this space.

The Data Point Rulesets can be defined in terms of Value Domains in order to maximize their reusability, in fact this way a Ruleset can be applied on any Data Set which has Components which take values on the Value Domains of the Ruleset signature. The association between the Components of the Data Set and the Value Domains of the Ruleset signature is provided by the **check_datapoint** operator at the invocation of the Ruleset.

When the Ruleset is defined on Variables, their reusability is intentionally limited to the Data Sets which contains such Variables (and not to other possible Variables which take values from the same Value Domain). If at a later stage the Ruleset would need to be applied also to other Variables defined on the same Value Domain, a similar Ruleset should be defined also for the other Variable.

Rules are uniquely identified by **ruleName**. If omitted then **ruleName** is implicitly assumed to be the progressive order number of the Rule in the Ruleset. Please note however that, using this default mechanism, the Rule Name can change if the Ruleset is modified, e.g., if new Rules are added or existing Rules are deleted, and therefore the users that interpret the validation results must be aware of these changes. In addition, if the results of more than one Ruleset have to be combined in one Data Set, then the user should make the relevant rulesetNames different.

As said, each Rule is applied in a row-wise fashion to each individual Data Point of a Data Set. The references to the Value Domains defined in the **antecedentCondition** and **consequentCondition** are replaced with the values of the respective Components of the Data Point under evaluation.

### Examples

```plaintext
define datapoint ruleset DPR_1 ( valuedomain flow_type as A, numeric_value as B ) is
  when A = "CREDIT" or A = "DEBIT" then B >= 0 errorcode "Bad value" errorlevel 10
end datapoint ruleset

define datapoint ruleset DPR_2 ( variable flow as F, obs_value as O ) is
  when F = "CREDIT" or F = "DEBIT" then O >= 0 errorcode "Bad value"
end datapoint ruleset
```

### define hierarchical ruleset

**Semantics**

This operator defines a persistent Hierarchical Ruleset that contains Rules to be applied to individual Components of a given Data Set in order to make validations or calculations according to hierarchical
relationships between the relevant Code Items. These Rule sets are also called "vertical" taking into account the 
tabular representation of a Data Set (considered as a mathematical function), in which each (vertical) column 
represents a variable and each (horizontal) row represents a Data Point: these Rule sets are applied on variables 
(columns), i.e., vertically on the tabular representation of a Data Set.
A main purpose of the hierarchical Rules is to express some more aggregated Code Items (e.g. the continents) in 
terms of less aggregated ones (e.g., their countries) by using Code Item Relationships. This kind of relations can 
be applied to aggregate data, for example to calculate an additive measure (e.g., the population) for the 
aggregated Code Items (e.g., the continents) as the sum of the corresponding measures of the less aggregated 
one(s), (e.g., their countries). These rules can be used also for validation, for example to check if the additive 
measures relevant to the aggregated Code Items (e.g., the continents) match the sum of the corresponding 
measures of their component Code Items (e.g., their countries), provided that the input Data Set contains all of 
them, i.e. the more and the less aggregated Code Items.
Another purpose of these Rules is to express the relationships in which a Code Item represents some part of 
another one, (e.g., "Africa" and "Five largest countries of Africa", being the latter a detail of the former). This kind 
of relationships can be used only for validation, for example to check if a positive and additive measure (e.g., the 
population) relevant to the more aggregated Code Item (e.g., Africa) is greater than the corresponding measure 
of the other more detailed one (e.g., "5 largest countries of Africa").
The name "hierarchical" comes from the fact that this kind of Rule set is able to express the hierarchical 
relationships between Code Items at different levels of detail, in which each (aggregated) Code Item is expressed 
as a partition of (disaggregated) ones. These relationships can be recursive, i.e., the aggregated Code Items can 
be in their turn component of even more aggregated ones, without limitations about the number of recursions.
As a first simple example, the following Hierarchical Ruleset named "BeneluxCountriesHierarchy" contains a 
single rule that asserts that, in the Value Domain "Geo_Area", the Code Item BENELUX is the aggregation of the 
Code Items BELGIUM, LUXEMBOURG and NETHERLANDS:

```
define hierarchical ruleset BeneluxCountriesHierarchy (valuedomain rule Geo_Area ) is
  BENELUX = BELGIUM + LUXEMBOURG + NETHERLANDS
end hierarchical ruleset
```

**Syntax**

```
define hierarchical ruleset  rulessetName ( hrRulesetSignature ) is
  hrRule
  { ; hrRule }*
end hierarchical ruleset
```

```
hrRulesetSignature ::= valuedomain { condition vdConditioningSignature } rule ruleValueDomain
```

```
vdConditioningSignature ::= condValueDomain { as vdAlias } { , condValueDomain { as vdAlias } }*
```

```
varRulesetSignature ::= variable { condition varConditioningSignature } rule ruleVariable
```

```
varConditioningSignature ::= condVariable { as vdAlias } { , condVariable { as vdAlias } }*
```

```
hrRule ::= { ruleName : } codeItemRelation { errorCode errorCode } { errorlevel errorLevel }
```

```
  codeItemRelation ::= 
    { when leftCondition then }
    leftCodeItem { = | > | < | >= | <= }*
    { + | - } rightCodeItem { [ [ rightCondition ] ] }
    { { + | - } } rightCodeItem { [ [ rightCondition ] ] })*
```

**Syntax description**

```
rulessetName                     the name of the Hierarchical Ruleset to be defined.
hrRulesetSignature               the signature of the Rule set. It specifies the Value Domain or Variable on which the
                                 Rule set is defined, and the Conditioning Signature.
vdRulesetSignature               the signature of a Rule set defined on Value Domains
varRulesetSignature             the signature of a Rule set defined on Variables
hrRule                           a single hierarchical rule, as described below.
vdConditioningSignature         specifies the Value Domains on which the conditions are defined. The Rule set is meant 
                                 to be applicable to the Data Sets having Components that take values on the Value
A variable on which the ruleset is defined (i.e., ruleValueDomain) and on the
conditioning Value Domains (i.e., condValueDomain).

The Value Domain on which the Ruleset is defined

an (optional) alias assigned to a Value Domain and valid only within the Ruleset, this
can be used for the sake of compactness in writing leftCondition and rightCondition. If an
alias is not specified then the name of the Value Domain (i.e., condValueDomain)
must be used.

The signature of the (possible) conditions of the Ruleset defined on Variables. It
specifies the Represented Variables (see the information model) on which these
conditions are defined. The Ruleset is meant to be applicable to any Data Set having
Components which are defined by the Variable on which the Ruleset is expressed (i.e.,,
variable) and on the Conditioning Variables.

the variable on which the Ruleset is defined

a conditioning Variable of the Ruleset

an (optional) alias assigned to a Variable and valid only within the Ruleset, this can be
used for the sake of compactness in writing leftCondition and rightCondition. If an
alias is not specified then the name of the Variable (parameter condVariable) must be used.

the name assigned to the specific Rule within the Ruleset. If the Ruleset is used for
validation then the ruleName identifies the validation results of the various Rules of
the Ruleset. The ruleName is optional and, if not specified, is assumed to be the
progressive order number of the Rule in the Ruleset. However please note that, if
ruleName is omitted then the Rule names can change in case the Ruleset is modified,
e.g., if new Rules are added or existing Rules are deleted, and therefore the users that
interpret the validation results must be aware of these changes. In addition, if the
results of more than one Ruleset have to be combined in one Data Set, then the user
should make the relevant rulesetNames different.

specifies a (possibly conditioned) Code Item Relation. It expresses a logical relation
between Code Items belonging to the Value Domain of the hrRulesetSignature, possibly conditioned by the Values of the Value Domains or Variables of the Conditioning Signature. The relation is expressed by one of the symbols =, >, >=, <, <=,
that in this context denote special logical relationships typical of Code Items. The first
member of the relation is a single Code Item. The second member of the relationship
is the composition of one or more Code Items combined using the symbols + or −,
which in turn also denote special logical operators typical of Code Items. The meaning
of these symbols is better explained below and in the User Manual.

a literal denoting the error code associated to the rule, to be assigned to the possible
non-valid results in case the Rule is used for validation. If omitted then no error code
is assigned (NULL value). VTL assumes that a Value Domain errorcode_vd of the error
codes exists in the Information Model and contains all the possible error codes: the
errorCode literal must be one of the possible Values of such a Value Domain. VTL
assumes also that a Variable errorcode for describing the error codes exists in the IM
and is a dependent variable of the Data Sets which contain the results of the
validation.

a literal denoting the error level (severity) associated to the rule, to be assigned to the
possible non-valid results in case the Rule is used for validation. If omitted then no error level is assigned (NULL value). VTL assumes that a Value Domain errorlevel_vd of the error levels exists in the Information Model and contains all the possible error levels: the errorLevel literal must be one of the possible Values of such a Value Domain. VTL assumes also that a Variable errorlevel for describing the error levels exists in the IM and is a dependent variable of the Data Sets which contain the results of the
validation.

a boolean expression which defines the pre-condition for evaluating the left member
Code Item (i.e., it is evaluated only when the leftCondition is TRUE); It can contain
references to the Value domains or the Variables of the conditioningSignature of the
Ruleset and Constants; all the VTL-ML component level operators are allowed. The
leftCondition is optional, if missing it is assumed to be TRUE and the Rule is always
evaluated.

a Code Item of the Value Domain specified in the hrRulesetSignature.
rightCodeItem: a Code Item of the Value Domain specified in the hrRulesetSignature.
rightCondition: a boolean scalar expression which defines the condition for a right member Code Item
to contribute to the evaluation of the Rule (i.e., the right member Code Item is taken
into account only when the relevant rightCondition is TRUE). It can contain references
to the Value Domains or Variables of the vdConditioningSignature or
varConditioningSignature of the Ruleset and Constants; all the VTL-ML component
level operators are allowed. The rightCondition is optional, if omitted then it is
assumed to be TRUE and the right member Code Item is always taken into account.

Input parameters type

rulesetName :: name < ruleset >
ruleValueDomain :: name < valuedomain >
condValueDomain :: name < valuedomain >
vAlias :: name
ruleVariable :: name
condVariable :: name
varAlias :: name
ruleName :: name
codeError :: errocode_vd
codeError :: errorlevel_vd
leftCondition :: boolean
leftCodeItem :: name
rightCodeItem :: name
rightCondition :: boolean

Constraints

- leftCondition and rightCondition can refer only to Value Domains or Variables specified in
  vdConditioningSignature or varConditioningSignature.
- Either the ruleName is specified for all the Rules of the Ruleset or for none.
- If specified, the ruleName must be unique within the Ruleset.

Semantic specification

This operator defines a Hierarchical Ruleset named rulesetName that can be used both for validation and
calculation purposes (see check_hierarchy and hierarchy). A Hierarchical Ruleset is a set of Rules expressing
logical relationships between the Values (Code Items) of a Value Domain or a Represented Variable.
Each rule contains a Code Item Relation, possibly conditioned, which expresses the relation between Code
Items to be enforced. In the relation, the left member Code Item is put in relation to a combination of one or
more right member Code Items. The kinds of relations are described below.
The left member Code Item can be optionally conditioned through a leftCondition, a boolean expression which
defines the cases in which the Rule has to be applied (if not declared the Rule is applied ever). The participation
of each right member Code Item in the Relation can be optionally conditioned through a rightCondition, a
boolean expression which defines the cases in which the Code Item participates in the relation (if not declared
the Code Item participates to the relation ever).
As for the mathematical meaning of the relation, please note that each Value (Code Item) is the representation of
an event belonging to a space of events (i.e., the relevant Value Domain), according to the notions of “event” and
“space of events” of the probability theory (see also the section on the Generic Models for Variables and Value
Domains in the VTL IM). Therefore the relations between Values (Code Items) express logical implications
between events.
The envisaged types of relations are: “coincides” (\(\equiv\)), “implies” (\(\leq\)), “implies or coincides” (\(\leq\equiv\)), “is implied by”
(\(\geq\)), “is implied by or coincides” (\(\geq\equiv\)). For example:
UnitedKingdom \(\leq\) Europe
means that UnitedKingdom implies Europe (if a point belongs to United Kingdom it also belongs to Europe).
January2000 \(\leq\) year2000
means that January of the year 2000 implies the year 2000 (if a time instant belongs to “January 2000” it also
belongs to the “year 2000”).
The first member of a Relation is a single Code Item. The second member can be either a single Code Item, like in
the example above, or a logical composition of Code Items giving another Code Item as result. The logical

\(^2\) “Coincides” means “implies and is implied”
composition can be defined by means of Code Item Operators, whose goal is to compose some Code Items in order to obtain another Code Item.

Please note that the symbols + and - do not denote the usual operations of sum and subtraction, but logical operations between Code Items which are seen as events of the probability theory. In other words, two or more Code Items cannot be summed or subtracted to obtain another Code Item, because they are events and not numbers, however they can be manipulated through logical operations like "OR" and "Complement".

Note also that the + also acts as a declaration that all the Code Items denoted by + in the formula are mutually exclusive one another (i.e., the corresponding events cannot happen at the same time), as well as the - acts as a declaration that all the Code Items denoted by - in the formula are mutually exclusive one another and furthermore that each one of them is a part of (implies) the result of the composition of all the Code Items having the + sign.

At intuitive level, the symbol + means "with" (Benelux = Belgium with Luxembourg with Netherlands) while the symbol - means "without" (EUwithoutUK = EuropeanUnion without UnitedKingdom).

When these relationships are applied to additive numeric measures (e.g., the population relevant to geographical areas), they allow to obtain the measure values of the compound Code Items (i.e., the population of Benelux and Netherlands). This is why these logical operations are denoted in VTL through the same symbols as the usual sum and subtraction. Please note also that this property is valid whichever is the Data Set and whichever is the additive measure (provided that the possible other Identifier Components of the Data Set Structure have the same values), therefore the Rulesets of this kind are potentially largely reusable.

The Ruleset Signature specifies the space on which the Ruleset is defined, i.e., the ValueDomain or Variable on which the Code Item Relations are defined (the Ruleset is meant to be applicable to Data Sets having a Component which takes values on such a Value Domain or are defined by such a Variable). The optional vdConditioningSignature specifies the conditioning Value Domains (the conditions can refer only to those Value Domains), as well as the optional varConditioningSignature specifies the conditioning Variables (the conditions can refer only to those Variables).

The Hierarchical Ruleset may act on one or more Measures of the input Data Set provided that these measures are additive (for example it cannot be applied on a measure containing a "mean" because it is not additive).

Within the Hierarchical Rulesets there can be dependencies between Rules, because the inputs of some Rules can be the output of other Rules, so the former can be evaluated only after the latter. For example, the data relevant to the Continents can be calculated only after the calculation of the data relevant to the Countries. As a consequence, the order of calculation of the Rules is determined by their mutual dependencies and can be different from the order in which the Rules are written in the Ruleset. The dependencies between the Rules form a directed acyclic graph.

The hierarchical ruleset can be used for calculations to calculate the upper levels of the hierarchy if the data relevant to the leaves (or some other intermediate level) are available in the operand Data Set of the hierarchy operator (for more information see also the "Hierarchy" operator). For example, having additive Measures broken by region, it would be possible to calculate these Measures broken by countries, continents and the world. Besides, having additive Measures broken by country, it would be possible to calculate the same Measures broken by continents and the world.

When a Hierarchical Ruleset is used for calculation, only the Relations expressing coincidence (=) are evaluated (provided that the leftCondition is TRUE, and taking into account only right-side Code Items whose rightCondition is TRUE). The result Data Set will contain the compound Code Items (the left members of those relations) calculated from the component Code Items (the right member of those Relations), which are taken from the input Data Set (for more details about the evaluation options see the hierarchy operator). Moreover, the clauses typical of the validation are ignored (e.g., ErrorCode, ErrorLevel).

The Hierarchical Ruleset can be also used to filter the input Data Points. In fact if some Code Items are defined equal to themselves, the relevant Data Points are brought in the result unchanged. For example, the following Ruleset will maintain in the result the Data Points of the input Data Set relevant to Belgium, Luxembourg and Netherlands and will add new Data Points containing the calculated value for Benelux:

```
define hierarchical ruleset BeneluxRuleset ( valuedomain rule GeoArea) is
  Belgium = Belgium
  ; Luxembourg = Luxembourg
  ; Netherlands = Netherlands
  ; Benelux = Belgium + Luxembourg + Netherlands
end hierarchical ruleset
```

The Hierarchical Rulesets can be used for validation in case various levels of detail are contained in the Data Set to be validated (see also the check_hierarchy operator for more details). The Hierarchical Rulesets express
the coherency Rules between the different levels of detail. Because in the validation the various Rules can be
evaluated independently, their order is not significant.

If a Hierarchical Ruleset is used for validation, all the possible Relations (\(=\), \(>\), \(\geq\), \(<\), \(\leq\)) are evaluated (provided
that the leftCondition is TRUE and taking into account only right-side Code Items whose rightCondition is TRUE).
The Rules are evaluated independently. Both the Code Items of the left and right members of the Relations are
expected to belong to and taken from the input Data Set (for more details about the evaluation options see the
check_hierarchy operator). The Antecedent Condition is evaluated and, if TRUE, the operations specified in the
right member of the Relation are performed and the result is compared to the first member, according to the
specified type of Relation. The possible relations in which Code Items are defined as equal to themselves are
ignored. Further details are described in the check_hierarchy operator.

If the data to be validated are in different Data Sets, either they can be joined in advance using the proper VTL
operators or the validation can be done by comparing those Data Sets directly, without using a Hierarchical
Ruleset (see also the check operator).

Through the right and left Conditions, the Hierarchical Rulesets allow to declare the time validity of
Rules and Relations. In fact leftCondition and RightCondition can be defined in term of the time Value Domain,
expressing respectively when the left member Code Item has to be evaluated (i.e., when it is considered valid)
and when a right member Code Item participates in the relation.
The following two simplified examples show possible ways of defining the European Union in term of
participating Countries.

**Example 1** (for simplicity the time literals are written without the needed "cast" operation)

```vrl
define hierarchical ruleset EuropeanUnionAreaCountries1
    (valuedomain condition ReferenceTime as Time rule Geo_Area)
        is
        World = Africa + America + Asia + Europe + Oceania
        ; Africa = Algeria + ... + Zimbabwe
```

**Example 2** (for simplicity the time literals are written without the needed "cast" operation)

```vrl
define hierarchical ruleset EuropeanUnionAreaCountries2
    (valuedomain condition ReferenceTime as Time rule Geo_Area)
        is
        EU = AT [ Time >= "0101.1995" ]
            + BE [ Time >= "01.01.1958" ]
            + BG [ Time >= "01.01.2007" ]
            + ...  
            + SE [ Time >= "01.01.1995" ]
            + SI [ Time >= "01.05.2004" ]
            + SK [ Time >= "01.05.2004" ]
```

The Hierarchical Rulesets allow defining hierarchies either having or not having levels (free hierarchies).
For example, leaving aside the time validity for sake of simplicity:

```vrl
define hierarchical ruleset GeoHierarchy (valuedomain rule Geo_Area)
    is
    World = Africa + America + Asia + Europe + Oceania
    ; Africa = Algeria + ... + Zimbabwe
```
The Hierarchical Rulesets allow defining multiple relations for the same Code Item.

Multiple relations are often useful for validation. For example, the Balance of Payments item “Transport” can be broken down both by type of carrier (Air transport, Sea transport, Land transport) and by type of objects transported (Passengers and Freight). In the following example a RuleName is assigned to the different methods of breaking down the Transport.

\[
\text{define hierarchical ruleset TransportBreakdown (variable rule BoPItem) is} \\
\text{transport\_method1 : Transport} = \text{AirTransport + SeaTransport + LandTransport} \\
\text{transport\_method2 : Transport} = \text{PassengersTransport + FreightTransport}
\]

The outcome of both breakdowns must sum up to the whole “Transport” figure. In the following example a RuleName is assigned to the different methods of breaking down the Transport.

Multiple relations can be useful even for calculation. For example, imagine that the input Data Set contains data about resident units broken down by region and data about non-residents units broken down by country. In order to calculate a homogeneous level of aggregation (e.g., by country), a possible Ruleset is the following:

\[
\text{define hierarchical ruleset CalcCountryLevel (valuedomain condition Residence rule GeoArea) is} \\
\text{when Residence} = \text{“resident” then Country1} = \text{Country1} \text{ + Region11 + … + Region1M} \\
\text{when Residence} = \text{“non-resident” then CountryN} = \text{CountryN} \text{ + RegionN1 + … + RegionNM}
\]

In the calculation, basically, for each Rule, for all the input Data Points and provided that the conditions are TRUE, the right Code Items are changed into the corresponding left Code Item, obtaining Data Points referred only to the left Code Items. Then the outcomes of all the Rules of the Ruleset are aggregated together to obtain the Data Points of the result Data Set.

As far as each left Code Item is calculated by means of a single Rule (i.e., a single calculation method), this process cannot generate inconsistencies. Instead if a left Code Item is calculated by means of more Rules (e.g., through more than one calculation method), there is the risk of producing erroneous results (e.g., duplicated data), because the outcome of the multiple Rules producing the same Code Item are aggregated together. Proper definition of the left or right conditions can avoid this risk, ensuring that for each input Data Point just one Rule is applied. If the Ruleset is aimed only at validation, there is no risk of producing erroneous results because in the validation the rules are applied independently.

Examples

1) The Hierarchical Rule set is defined on the Value Domain “sex”: Total is defined as Male + Female.

\[
\text{define hierarchical ruleset sex\_hr (valuedomain rule sex) is} \\
\text{TOTAL} = \text{MALE + FEMALE}
\]

2) BENELUX is the aggregation of the Code Items BELGIUM, LUXEMBOURG and NETHERLANDS. No conditions are defined.

\[
\text{define hierarchical ruleset BeneluxCountriesHierarchy (valuedomain rule GeoArea) is} \\
\text{BENELUX} = \text{BELGIUM} + \text{LUXEMBOURG} + \text{NETHERLANDS errorcode “Bad value for Benelux”}
\]
3) American economic partners. The first rule states that the value for North America should be greater than the value reported for US. This type of validation is useful when the data communicated by the data provider do not cover the whole composition of the aggregate but only some elements. No conditions are defined.

```plaintext
define hierarchical ruleset american_partners_hr (variable rule PartnerArea) is
  NORTH_AMERICA > US
  ; SOUTH_AMERICA = BR + UY + AR + CL
end hierarchical ruleset
```

4) Example of an aggregate Code Item having multiple definitions to be used for validation only. The Balance of Payments item "Transport" can be broken down by type of carrier (Air transport, Sea transport, Land transport) and by type of objects transported (Passengers and Freights) and both breakdowns must sum up to the total "Transport" figure.

```plaintext
define hierarchical ruleset validationruleset_bop (variable rule BoPItem ) is
  transport_method1 : Transport = AirTransport + SeaTransport + LandTransport
  ; transport_method2 : Transport = PassengersTransport + FreightsTransport
end hierarchical ruleset
```
define operator

Syntax

```vtl
define operator operator_name ( { parameter { , parameter }* } )
{ returns outputType }
is operatorBody
end operator
```

```vtl
parameter ::= parameterName parameterType { default parameterDefaultValue }
```

Syntax description

- `operator_name`: the name of the operator
- `parameter`: the names of parameters, their data types and default values
- `outputType`: the data type of the artefact returned by the operator
- `operatorBody`: the expression which defines the operation
- `parameterName`: the name of the parameter
- `parameterType`: the data type of the parameter
- `parameterDefaultValue`: the default value for the parameter (optional).

Parameters

- `operator_name`: name
- `outputType`: a VTL data type as defined in `outputParameterType` (see the Data Type Syntax)
- `operatorBody`: a VTL expression having the parameters (i.e., `parameterName`) as the operands
- `parameterName`: name
- `parameterType`: a VTL data type as defined in `inputParameterType` (see the Data Type Syntax)
- `parameterDefaultValue`: a Value of the same type as the parameter

Constraints

- Each `parameterName` must be unique within the list of parameters
- `parameterDefaultValue` must be of the same data type as the corresponding parameter
- If `outputType` is specified then the type of `operatorBody` must be compatible with `outputType`
- If `outputType` is omitted then the type returned by the `operatorBody` expression is assumed
- If `parameterDefaultValue` is specified then the parameter is optional

Semantic specification

This operator defines a user-defined Operator by means of a VTL expression, specifying also the parameters, their data types, whether they are mandatory or optional and their (possible) default values.

Examples

**Example1**:  
```vtl
define operator max1 (x integer, y integer)
returns boolean is
if x > y then x else y
end operator
```

**Example2**:  
```vtl
define operator add (x integer default 0, y integer default 0)
returns number is
x+y
end operator
```
Data type syntax

The VTL data types are described in the VTL User Manual. Types are used throughout this Reference Manual as both meta-syntax and syntax.

They are used as meta-syntax in order to define the types of input and output parameters in the descriptions of VTL operators; they are used in the syntax, and thus are proper part of the VTL, in order to allow other operators to refer to specific data types. For example, when defining a custom operator (see the define operator above), one will need to declare the type of the input/output parameters.

The syntax of the data types is described below (as for the meaning of these definitions, see the section VTL Data Types in the User Manual). See also the section “Conventions for describing the operators’ syntax” in the chapter “Overview of the language and conventions” above.

datatype ::= scalarType | scalarSetType | componentType | datasetType | operatorType | rulesetType

scalarType ::= { basicScalarType | valueDomainName | setName } { { scalarTypeConstraint } { not | null } }

basicScalarType ::= scalar | number | integer | string | boolean | time | date | time_period | duration

scalarTypeConstraint ::= [ valueBooleanCondition ] { { scalarLiteral | , scalarLiteral }* }

scalarSetType ::= set { < scalarType > }

componentType ::= componentRole { < scalarType > }

componentRole ::= component | identifier | measure | attribute | viral attribute

datasetType ::= dataset { { componentConstraint | , componentConstraint }* }

componentConstraint ::= componentType { componentName | multiplicityModifier }*

multiplicityModifier ::= _{ + | * }*

operatorType ::= inputParameterType { * inputParameterType }* -> outputParameterType

inputParameterType ::= scalarType | scalarSetType | componentType | datasetType | rulesetType

outputParameterType ::= scalarType | componentType | datasetType

rulesetType ::= ruleset | dpRuleset | hrRuleset

dpRuleset ::= datapoint

| datapoint_on_valuedomains { { valueDomainName | * valueDomainName }* } }

| datapoint_on_variables { { variableName | * variableName }* } }

hrRuleset ::= hierarchical

| hierarchical_on_valuedomains { { valueDomainName 

| ( { condValueDomainName | * condValueDomainName }* ) } }

| hierarchical_on_variables { { variableName 

| ( { condVariableName | * condVariableName }* ) } }

Note that the valueBooleanCondition in scalarTypeConstraint is expressed with reference to the fictitious variable “value” (see also the User Manual, section “Conventions for describing the Scalar Types”), which represents the generic value of the scalar type, for example:
General examples of the syntax for defining types can be found in the User Manual, section VTL Data Types and in the declaration of the data types of the VTL operators (sub-sections “input parameters type” and “result type”).
In this section, the common behaviours of some class of VTL-ML operators are described, both for a better understanding of the characteristics of such classes and to factor out and not repeat the explanation for each operator of the class.

Typical behaviour of most ML Operators

Unless differently specified in the Operator description, the Operators can be applied to Scalar Values, to Data Sets and to Data Set Components.

The operations on Scalar Values are primitive and are part of the core of the language. The other kind of operations can be typically be obtained by means of the scalar operations in conjunction with the Join operator, which is part of the core too.

In the operations on Data Sets, the Operators are meant to be applied by default only to the values of the Measures of the input Data Sets, leaving the Identifiers unchanged. The Attributes follow by default their specific propagation rules, which are described in the User Manual.

In the operations on Components, the Operators are meant to be applied on the specified components of one input Data Set, in order to calculate a new component which becomes part of the resulting Data Set. In this case, the Attributes can be operated like the Measures.

Operators applicable on one Scalar Value or Data Set or Data Set Component

Operations on Scalar Values

The operator is applied on a scalar value and returns a scalar value.

Operations on Data Sets

The operator is applied on a Data Set and returns a Data Set.

For example, using a functional style and denoting the operator with \( f(\ldots) \), this can written as:

\[
DS_r := f(DS_1)
\]

The same operation, using an infix style and denoting the operator as \( \text{op} \), can be also written as

\[
DS_r := \text{op} DS_1
\]

This means that the operator is applied to the values of all the Measures of DS_1 in order to produce homonymous Measures in DS_r.

The application of the operator is allowed only if all the Measures of the operand Data Set are of a data type compatible with the operator (for example, a numeric operator is applicable only if all the Measures of the operand Data Sets are numeric). If the Measures of the operand Data Set are of different types, not all compatible with the operator to be applied, the membership or the keep clauses can be used to select only the proper Measures. No applicability constraints exist on Identifiers and Attributes, which can be any.

As for the data content, for each Data Point (DP_1) of the operand Data Set, a result Data Point (DP_r) is returned, having for the Identifiers the same values as DP_1.

For each Data Point DP_1 and for each Measure, the operator is applied on the Measure value of DP_1 and returns the corresponding Measure value of DP_r.

For each Data Point DP_1 and for each viral Attribute, the value of the Attribute propagates unchanged in DP_r.

As for the data structure, the result Data Set (DS_r) has the Identifiers and the Measures of the operand Data Set (DS_1), and has the Attributes resulting from the application of the attribute propagation rules on the Attributes of the operand Data Set (DS_r maintains the Attributes declared as "viral" in DS_1; these Attributes are considered as "viral" also in DS_r, the "non-viral" Attributes of DS_1 are not kept in DS_r).
Operations on Data Set Components

The operator is applied on a Component (COMP_1) of a Data Set (DS_1) and returns another Component (COMP_r) which alters the structure of DS_1 in order to produce the result Data Set (DS_r).

For example, using a functional style and denoting the operator with \( f \), this can be written as:

\[
DS_r := DS_1 \ [\text{calc}\ COMP_r := f\ (COMP_1) ]
\]

The same operation, using an infix style and denoting the operator as \( \text{op} \), can be written as:

\[
DS_r := DS_1 \ [\text{calc}\ COMP_r := \text{op}\ COMP_1 ]
\]

This means that the operator is applied on COMP_1 in order to calculate COMP_r.

- If COMP_r is a new Component which originally did not exist in DS_1, it is added to the original Components of DS_1, by default as a Measure (unless otherwise specified), in order to produce DS_r.
- If COMP_r is one of the original Measures or Attributes of DS_1, the values obtained from the application of the operator \( f \) replace the DS_1 original values for such a Measure or Attribute in order to produce DS_r.
- If COMP_r is one of the original Identifiers of DS_1, the operation is not allowed, because the result can become inconsistent.

In any case, an operation on the Components of a Data Set produces a new Data Set, as in the example above.

The application of the operator is allowed only if the input Component belongs to a data type compatible with the operator (for example, a numeric operator is applicable only on numeric Components). As already said, COMP_r cannot have the same name of an Identifier of DS_1.

As for the data content, for each Data Point DP_1 of DS_1, the operator is applied on the values of COMP_1 so returning the value of COMP_r.

As for the data structure, like for the operations on Data Sets above, the result Data Set (DS_r) has the Identifiers and the Measures of the operand Data Set (DS_1), and has the Attributes resulting from the application of the attribute propagation rules on the Attributes of the operand Data Set (DS_r maintains the Attributes declared as "viral" in DS_1; these Attributes are considered as "viral" also in DS_r, the "non-viral" Attributes of DS_1 are not kept in DS_r). If an Attribute is explicitly calculated, the attribute propagation rule is overridden.

Moreover, in the case of the operations on Data Set Components, the (possibly) new Component DS_r can be added to the original structure, the role of a (possible) existing Component DS_r can be altered, the vitality of a (possibly) existing DS_r Attribute can be altered, a (possible) COMP_r non-viral Attribute can be kept in the result. For the alteration of role and virality see also the \texttt{calc} clause.

Operators applicable on two Scalar Values or Data Sets or Data Set Components

Operation on Scalar values

The operator is applied on two Scalar values and returns a Scalar value.

Operation on Data Sets

The operator is applied either on two Data Sets or on one Data Set and one Scalar value and returns a Data Set.

The composition of a Data Set and a Component is not allowed (it makes no sense).

For example, using a functional style and denoting the operator with \( f \), this can be written as:

\[
DS_r := f\ (DS_1, DS_2)
\]

The same kind of operation, using an infix style and denoting the operator as \( \text{op} \), can be also written as

\[
DS_r := DS_1 \ \text{op}\ DS_2
\]

This means that the operator is applied to the values of all the couples of Measures of DS_1 and DS_2 having the same names in order to produce homonymous Measures in DS_r. DS_1 or DS_2 may be replaced by a Scalar value.

The composition of two Data Sets (DS_1, DS_2) is allowed if the two operand Data Sets have exactly the same Measures and if all these Measures belong to a data type compatible with the operator (for example, a numeric operator is applicable only if all the Measures of the operand Data Sets are numeric). If the Measures of the operand Data Sets are different or of different types not all compatible with the operator to be applied, the membership or the \texttt{keep} clauses can be used to select only the proper Measures. The composition is allowed if
these operand Data Sets have the same Identifiers or if one of them has at least all the Identifiers of the other one
(in other words, the Identifiers of one of the Data Sets must be a superset of the Identifiers of the other one). No
applicability constraints exist on the Attributes, which can be any.
As for the data content, the operand Data Sets (DS_1, DS_2) are joined to find the couples of Data Points (DP_1, DP_2),
where DP_1 is from the first operand (DS_1) and DP_2 from the second operand (DS_2), which have the
same values as for the common Identifiers. Data Points that are not coupled are left out (the inner join is used).
An operand Scalar value is treated as a Data Point that couples with all the Data Points of the other operand. For
each couple (DP_1, DP_2) a result Data Point (DP_r) is returned, having for the Identifiers the same values as
DP_1 and DP_2.
For each Measure and for each couple (DP_1, DP_2), the Measure values of DP_1 and DP_2 are composed through
the operator so returning the Measure value of DP_r. An operand Scalar value is composed with all the Measures
of the other operand.
For each couple (DP_1, DP_2) and for each Attribute that propagates in DP_r, the Attribute value is calculated by
applying the proper Attribute propagation algorithm on the values of the Attributes of DP_1 and DP_2.
As for the data structure, the result Data Set (DS_r) has all the Identifiers (with no repetition of common
Identifiers) and the Measures of both the operand Data Sets, and has the Attributes resulting from the
application of the attribute propagation rules on the Attributes of the operands (DS_r maintains the Attributes
declared as "viral" for the operand Data Sets; these Attributes are considered as "viral" also in DS_r, the "non-
viral" Attributes of the operand Data Sets are not kept in DS_r).

Operation on Data Set Components

The operator is applied either on two Data Set Components (COMP_1, COMP_2) belonging to the same Data Set
(DS_1) or on a Component and a Scalar value, and returns another Component (COMP_r) which alters the
structure of DS_1 in order to produce the result Data Set (DS_r). The composition of a Data Set and a Component
is not allowed (it makes no sense).
For example, using a functional style and denoting the operator with \( f(\ldots) \), this can be written as:

\[
DS_r := DS_1 \left[ \text{calc} \ COMP_r := f(\ COMP_1, COMP_2 ) \right]
\]

The same operation, using an infix style and denoting the operator as \( \text{op} \), can be written as:

\[
DS_r := DS_1 \left[ \text{calc} \ COMP_r := COMP_1 \ \text{op} \ \ COMP_2 \right]
\]

This means that the operator is applied on COMP_1 and COMP_2 in order to calculate COMP_r.

- If COMP_r is a new Component which originally did not exist in DS_1, it is added to the original Components
  of DS_1, by default as a Measure (unless otherwise specified), in order to produce DS_r.
- If COMP_r is one of the original Measures or Attributes of DS_1, the values obtained from the application of
  the operator \( f(\ldots) \) replace the DS_1 original values for such a Measure or Attribute in order to produce
  DS_r.
- If COMP_r is one of the original Identifiers of DS_1, the operation is not allowed, because the result can
  become inconsistent.

In any case, an operation on the Components of a Data Set produces a new Data Set, like in the example above.
The composition of two Data Set Components is allowed provided that they belong to the same Data Set³. Moreover, the input Components must belong to data types compatible with the operator (for example, a numeric operator is applicable only on numeric Components). As already said, COMP_r cannot have the same
name of an Identifier of DS_1.
As for the data content, for each Data Point of DS_1, the values of COMP_1 and COMP_2 are composed through
the operator so returning the value of COMP_r.
As for the data structure, the result Data Set (DS_r) has the Identifiers and the Measures of the operand Data Set
(DS_1), and has the Attributes resulting from the application of the attribute propagation rules on the Attributes
of the operand Data Set (DS_r maintains the Attributes declared as "viral" in DS_1; these Attributes are
explicitly calculated, the attribute propagation rule is overridden.
Moreover, in the case of the operations on Data Set Components, a (possible) new Component DS_r can be added
to the original structure of DS_1, the role of a (possibly) existing DS_1 Component can be altered, the virality of a

³ As obvious, the input Data Set can be the result of a previous composition of more other Data Sets, even within the
same expression
(possibly) existing DS_r Attributes can be altered, a (possible) COMP_r non-viral Attribute can be kept in the result. For the alteration of role and virality see also the **calc** clause.

### Operators applicable on more than two Scalar Values or Data Set Components

The cases in which an operator can be applied on more than two Data Sets (like the Join operators) are described in the relevant sections.

#### Operation on Data Set Components

The operator is applied either on a combination of more than two Data Set Components (COMP_1, COMP_2) belonging to the same Data Set (DS_1) or Scalar values, and returns another Component (COMP_r) which alters the structure of DS_1 in order to produce the result Data Set (DS_r). The composition of a Data Set and a Component is not allowed (it makes no sense).

For example, using a functional style and denoting the operator with \( f( \ldots ) \), this can be written as:

\[
\text{DS}_r := \text{DS}_1 \left[ \text{substr} \text{COMP}_r := f(\text{COMP}_1, \text{COMP}_2, \text{COMP}_3) \right]
\]

This means that the operator is applied on COMP_1, COMP_2 and COMP_3 in order to calculate COMP_r.

- If COMP_r is a new Component which originally did not exist in DS_1, it is added to the original Components of DS_1, by default as a Measure (unless otherwise specified), in order to produce DS_r.
- If COMP_r is one of the original Measures or Attributes of DS_1, the values obtained from the application of the operator \( f( \ldots ) \) replace the DS_1 original values for such a Measure or Attribute in order to produce DS_r.
- If COMP_r is one of the original Identifiers of DS_1, the operation is not allowed, because the result can become inconsistent.

In any case, an operation on the Components of a Data Set produces a new Data Set, like in the example above.

The composition of more Data Set Components is allowed provided that they belong to the same Data Set\(^4\). Moreover, the input Components must belong to data types compatible with the operator (for example, a numeric operator is applicable only on numeric Components). As already said, COMP_r cannot have the same name of an Identifier of DS_1.

As for the data content, for each Data Point of DS_1, the values of COMP_1, COMP_2 and COMP_3 are composed through the operator so returning the value of COMP_r.

As for the data structure, the result Data Set (DS_r) has the Identifiers and the Measures of the operand Data Set (DS_1), and has the Attributes resulting from the application of the attribute propagation rules on the Attributes of the operand Data Set (DS_r maintains the Attributes declared as "viral" in DS_1; these Attributes are considered as "viral" also in DS_r, the "non-viral" Attributes of DS_1 are not kept in DS_r). If an Attribute is explicitly calculated, the attribute propagation rule is overridden.

Moreover, in the case of the operations on Data Set Components, a (possible) new Component DS_r can be added to the original structure of DS_1, the role of a (possibly) existing DS_1 Component can be altered, the virality of a (possibly) existing DS_r Attributes can be altered, a (possible) COMP_r non-viral Attribute can be kept in the result. For the alteration of role and virality see also the **calc** clause.

#### Behaviour of Boolean operators

The Boolean operators are allowed only on operand Data Sets that have a single measure of type **boolean**. As for the other aspects, the behaviour is the same as the operators applicable on one or two Data Sets described above.

---

\(^4\) As obvious, the input Data Set can be the result of a previous composition of more other Data Sets, even within the same expression.
Behaviour of Set operators

These operators apply the classical set operations (union, intersection, difference, symmetric differences) to the Data Sets, considering them as sets of Data Points. These operations are possible only if the Data Sets to be operated have the same data structure, and therefore the same Identifiers, Measures and Attributes.  

Behaviour of Time operators

The time operators are the operators dealing with time, date and time_period basic scalar types. These types are described in the User Manual in the sections “Basic Scalar Types” and “External representations and literals used in the VTL Manuals”.

The time-related formats used for explaining the time operators are the following (they are described also in the User Manual). 

For the time values:

\[ YYYY-MM-DD/YYY-MM-DD \]

Where YYYY are 4 digits for the year, MM two digits for the month, DD two digits for the day. For example:

- 2000-01-01/2000-12-31 the whole year 2000
- 2000-01-01/2009-12-31 the first decade of the XXI century

For the date values:

\[ YYYY-MM-DD \]

The meaning of the symbols is the same as above. For example:

- 2000-12-31 the 31st December of the year 2000
- 2010-01-01 the first of January of the year 2010

For the time_period values:

\[ YYYY(P)NNN \]

Where YYYY are 4 digits for the year, P is one character for the period indicator of the regular period (it refers to the duration data type and can assume one of the possible values listed below), NNN are from zero to three digits which contain the progressive number of the period in the year. For annual data the A and the three digits NNN can be omitted. For example:

- 2000M12 the month of December of the year 2000 (duration: M)
- 2010Q1 the first quarter of the year 2010 (duration: Q)
- 2010A the whole year 2010 (duration: A)
- 2010 the whole year 2010 (duration: A)

For the duration values, which are the possible values of the period indicator of the regular periods above, it is used for simplicity just one character whose possible values are the following:

<table>
<thead>
<tr>
<th>Code</th>
<th>Duration</th>
</tr>
</thead>
<tbody>
<tr>
<td>D</td>
<td>Day</td>
</tr>
<tr>
<td>W</td>
<td>Week</td>
</tr>
<tr>
<td>M</td>
<td>Month</td>
</tr>
<tr>
<td>Q</td>
<td>Quarter</td>
</tr>
<tr>
<td>S</td>
<td>Semester</td>
</tr>
<tr>
<td>A</td>
<td>Year</td>
</tr>
</tbody>
</table>

As mentioned in the User Manual, these are only examples of possible time-related representations, each VTL system is free of adopting different ones. In fact no predefined representations are prescribed, VTL systems are free to using they preferred or already existing ones.

Several time operators deal with the specific case of Data Sets of time series, having an Identifier component that acts as the reference time and can be of one of the scalar types time, date or time_period; moreover this Identifier must be periodical, i.e. its possible values are regularly spaced and therefore have constant duration (frequency). 

5 According to the VTL IM, the Variables that have the same name have also the same data type
It is worthwhile to recall here that, in the case of Data Sets of time series, VTL assumes that the information about which is the Identifier Components that acts as the reference time and which is the period (frequency) of the time series exists and is available in some way in the VTL system. The VTL Operators are aware of which is the reference time and the period (frequency) of the time series and use these information to perform correct operations. VTL also assumes that a Value Domain representing the possible periods (e.g. the period indicator Value Domain shown above) exists and refers to the duration scalar type. For the assumptions above, the users do not need to specify which is the Identifier Component having the role of reference time.

The operators for time series can be applied only on Data Sets of time series and returns a Data Set of time series. The result Data Set has the same Identifier, Measure and Attribute Components as the operand Data Set and contains the same time series as the operand. The Attribute propagation rule is not applied.

### Operators changing the data type

These Operators change the Scalar data type of the operands they are applied to (i.e. the type of the result is different from the type of the operand). For example, the length operator is applied to a value of string type and returns a value of integer type. Another example is the cast operator.

#### Operation on Scalar values

The operator is applied on (one or more) Scalar values and returns one Scalar value of a different data type.

#### Operation on Data Sets

If an Operator changes the data type of the Variable it is applied to (e.g., from string to number), the result Data Set cannot maintain this Variable as it happens in the previous cases, because a Variable cannot have different data types in different Data Sets. As a consequence, the converted variable cannot follow the same rules described in the sections above and must be replaced, in the result Data Set, by another Variable of the proper data type.

For sake of simplicity, the operators changing the data type are allowed only on mono-measure operand Data Sets, so that the conversion happens on just one Measure. A default generic Measure is assigned by default to the result Data Set, depending on the data type of the result (the default Measure Variables are reported in the table below). Therefore, if the operands are originally multi-measure, just one Measure must be pre-emptively selected (for example through the membership operator) in order to apply the changing-type operator. Moreover, if in the result Data Set a different Measure Variable name is desired than the one assigned by default, it is possible to change the Variable name (see the rename operator).

As for the Identifiers and the Attributes, the behaviour of these operators is the same as the typical behaviour of the unary or binary operators.

#### Operation on Data Set Components

For the same reasons above, the result Component cannot be the same as one of the operand Components and must be of the appropriate Scalar data type.

### Default Names for Variables and Value Domains used in this manual

The following table shows the default Variable names and the relevant default Value Domain. These are only the names used in this manual for explanatory purposes and can be personalised in the implementations. If VTL rules are exchanged, the personalised names need to be shared with the partners of the exchange.

<table>
<thead>
<tr>
<th>Scalar data type</th>
<th>Default Variable</th>
<th>Default Value Domain</th>
</tr>
</thead>
<tbody>
<tr>
<td>string</td>
<td>string_var</td>
<td>string_vd</td>
</tr>
</tbody>
</table>

6 This according both to the mathematical meaning of a Variable and the VTL Information Model; in fact a Represented Variable is defined on just one Value Domain, which has just one data type, independently of the Data Structures and the Data Sets in which the Variable is used.
Type Conversion and Formatting Mask

The conversions between scalar types is provided by the operator `cast`, described in the section of the general purpose operators. Some particular types of conversion require the specification of a formatting mask, which specifies which format the source or the destination of the conversion should assume. The formatting masks for the various scalar types are explained here.

If needed, the formatting masks can be personalized in the VTL implementations. If VTL rules are exchanged, the personalized masks need to be shared with the partners of the exchange.

### The Numbers Formatting Mask

The `number` formatting mask can be defined as a combination of characters whose meaning is the following:

- "D" one numeric digit (if the scientific notation is adopted, D is only for the mantissa)
- "E" one numeric digit (for the exponent of the scientific notation)
- "*" an arbitrary number of digits
- "+" at least one digit
- "." (dot) can be used as a separator between the integer and the decimal parts.
- "," (comma) can be used as a separator between the integer and the decimal parts.

Examples of valid masks are:

```
DD.DDDDD, DD.D, D, D.DDDD, D*.D*, D+.D+ , DD.DDEEEE
```

### The Time Formatting Mask

The format of the values of the types `time`, `date` and `time_period` can be specified through specific formatting masks. A mask related to `time`, `date` and `time_period` is formed by a sequence of symbols which denote:

- the time units that are used, for example years, months, days
- the format in which they are represented, for example 4 digits for the year (2018), 2 digits for the month within the year (04 for April) and 2 digits for the day within the year and the month (05 for the 5th)
- the order of these parts; for example, first the 4 digits for the year, then the 2 digits for the month and finally the 2 digits for the day
- other (possible) typographical characters used in the representation; for example, a line between the year and the month and between the month and the day (e.g., 2018-04-05).

The time formatting masks follow the general rules below.

For a numerical representations of the time units:

- A digit is denoted through the use of a special character which depends on the time unit, for example Y is for "year", M is for "month" and D is for "day"
- The special character is lowercase for the time units shorter than the day (for example h for "hour", m for "minute", s for "second") and uppercase for time units equal to "day" or longer (for example W for "week", Q for "quarter", S for "semester")
The number of letters matches the number of digits, for example YYYY means that the year is represented with four digits and MM that the month is of 2 digits.

The numerical representation is assumed to be padded by leading 0 by default, for example MM means that April is represented as 04 and the year 33 AD as 0033.

If the numerical representation is not padded, the optional digits that can be omitted (if equal to zero) are enclosed within braces; for example {M}M means that April is represented by 4 and December by 12, while {YYY}Y means that the 33 AD is represented by 33.

For textual representations of the time units:

- **Special words** denote a textual localized representation of a certain unit, for example DAY means a textual representation of the day (MONDAY, TUESDAY ...).
- An optional number following the special word denote the maximum length, for example DAY3 is a textual representation that uses three characters (MON, TUE ...).
- The case of the special word correspond to the case of the value; for example day3 (lowercase) denotes the values mon, tue ...
- The case of the initial character of the special word correspond to the case of the initial character of the time format; for example Day3 denotes the values Mon, Tue ...
- The letter P denotes the period indicator, (i.e., day, week, month ...) and the letter p denotes ond digit for the number of periods.

Representation of more time units:

- If more time units are used in the same mask (for example years, months, days), it is assumed that the more detailed units (e.g., the day) are expressed through the order number that they assume within the less detailed ones (e.g., the month and the year). For example, if years, weeks and days are used, the weeks are within the year (from 1 to 53) and the days are within the year and the week (from 1 to 7).
- The position of the digits in the mask denotes the position of the corresponding values; for example, YYYYMMDD means four digits for the year followed by two digits for the month and then two digits for the day (e.g., 20180405 means the year 2018, month April, day 5th).
- Any other character can be used in the mask, meaning simply that it appears in the same position; for example, YYYY-MM-DD means that the values of year, month and day are separated by a line (e.g., 2018-04-05 means the year 2018, month April, day 5th) and \PMM denotes the letter “P” followed by two characters for the month.
- The special characters and the special words, if prefixed by the reverse slash (\) in the mask, appear in the same position in the time format; for example \PMM\M means the letter “P” followed by two characters for the month and then the letter “M”; for example, \P03M means a period of three months (this is an ISO 8601 standard representation for a period of MM months). The reverse slash can appear in the format if needed by prefixing it with another reverse slash; for example YYYY\MM means for digits for the year, a backslash and two digits for the month.

- The **special characters** and the corresponding time units are the following:
  - C century
  - Y year
  - S semester
  - Q quarter
  - M month
  - W week
  - D day
  - h hour digit (by default on 24 hours)
  - m minute
  - s second
  - d decimal of second
  - P period indicator (see the “duration” codes below)
  - p number of periods

The **special words** for textual representations are the following:
Examples of formatting masks for the time scalar type:

A Scalar Value of type time denotes time intervals of any duration and expressed with any precision, which are the intervening time between two time points.

These examples are about three possible ISO 8601 formats for expressing time intervals:

- Start and end time points, such as "2015-03-03T09:30:45Z/2018-04-05T12:30:15Z"
  
  VTL Mask: YYYY-MM-DDThh:mm:ssZ/YYYY-MM-DDThh:mm:ssZ

- Start and duration, such as "2015-03-03T09:30:45-01/P1Y2M10DT2H30M"
  
  VTL Mask: YYYY-MM-DDThh:mm:ss-01/PY\YM\MDD\DT\h\Hmm\M

- Duration and end, such as "P1Y2M10DT2H30M/2018-04-05T12:30:00+02"
  
  VTL Mask: PY\YM\MDD\DT/YYYY-MM-DDThh:mm:ssZ

Example of other possible ISO formats having accuracy reduced to the day

- Start and end, such as "20150303/20180405"
  
  VTL Mask: YYYY-MM-DD/YYYY-MM-DD

- Start and duration, such as "2015-03-03/P1Y2M10D"
  
  VTL Mask: YYYY-MM-DD/PY\YM\MDD\D

- Duration and end, such as "P1Y2M10D/2018-04-05"
  
  VTL Mask: PY\YM\MDD\DT/YYYY-MM-DD

Examples of formatting masks for the date scalar type:

A date scalar type is a point in time, equivalent to an interval of time having coincident start and end duration equal to zero.

These examples about possible ISO 8601 formats for expressing dates:

- Date and day time with separators: "2015-03-03T09:30:45Z"
  
  VTL Mask: YYYY-MM-DDThh:mm:ssZ

- Date and day time without separators "20150303T093045-01 "
  
  VTL Mask: YYYYMMDDThhmmss-01

Example of other possible ISO formats having accuracy reduced to the day

- Date and day-time with separators "2015-03-03/2018-04-05"
  
  VTL Mask: YYYY-MM-DD/YYYY-MM-DD

- Start and duration, such as "2015-03-03/P1Y2M10D"
  
  VTL Mask: YYYY-MM-DD/PY\YM\MDD\D

Examples of formatting masks for the time_period scalar type:

A time_period denotes non-overlapping time intervals having a regular duration (for example the years, the quarters of years, the months, the weeks and so on). The time_period values include the representation of the duration of the period.

These examples are about possible formats for expressing time-periods:

- Generic time period within the year such as: "2015Q4", "2015M12""2015D365"
  
  VTL Mask: YYYY{ppp} where P is the period indicator and ppp three digits for the number of periods, in the values, the period indicator may assume one of the values of the duration scalar type listed below.

- Monthly period: "2015M03"
  
  VTL Mask: YYYY\MM
Examples of formatting masks for the duration scalar type:

A Scalar Value of type duration denotes the length of a time interval expressed with any precision and without connection to any particular time point (for example one year, half month, one hour and fifteen minutes).

These examples are about possible formats for expressing durations (period / frequency)

• Non ISO representation of the duration in one character, whose possible codes are:

<table>
<thead>
<tr>
<th>Code</th>
<th>Duration</th>
</tr>
</thead>
<tbody>
<tr>
<td>D</td>
<td>Day</td>
</tr>
<tr>
<td>W</td>
<td>Week</td>
</tr>
<tr>
<td>M</td>
<td>Month</td>
</tr>
<tr>
<td>Q</td>
<td>Quarter</td>
</tr>
<tr>
<td>S</td>
<td>Semester</td>
</tr>
<tr>
<td>A</td>
<td>Year</td>
</tr>
</tbody>
</table>

VTL Mask: P (period indicator)

• ISO 8601 composite duration: "P10Y2M12DT02H30M15S" (P stands for "period")
VTL Mask: \PYY\YM\MDD\DTTh\Hmm\Mss\S

• ISO 8601 duration in weeks: "P018W" (P stands for "period")
VTL Mask: \PWWW\W

• ISO 4 characters representation: P10M (ten months), P02Q (two quarters) ...
VTL Mask: \PppP

Examples of fixed characters used in the ISO 8601 standard which can appear as fixed characters in the relevant masks:

P    designator of duration
T    designator of time
Z    designator of UTC zone
+    designator of offset from UTC zone
-    designator of offset from UTC zone
/    time interval separator

Attribute propagation

The VTL has different default behaviours for Attributes and for Measures, to comply as much as possible with the relevant manipulation needs. At the Data Set level, the VTL Operators manipulate by default only the Measures and not the Attributes. At the Component level, instead, Attributes are calculated like Measures, therefore the algorithms for calculating Attributes, if any, can be specified explicitly in the invocation of the Operators. This is the behaviour of clauses like calc, keep, drop, rename and so on, either inside or outside the join (see the detailed description of these operators in the Reference Manual).

The users which want to automatize the propagation of the Attributes’ Values can optionally enforce a mechanism, called Attribute Propagation rule, whose behaviour is explained in the User Manual (see the section “Behaviour for Attribute Components”). The adoption of this mechanism is optional, users are free to allow the attribute propagation rule or not. The users that do not want to allow Attribute propagation rules simply will not implement what follows.

In short, the automatic propagation of an Attribute depends on a Boolean characteristic, called “virality”, which can be assigned to any Attribute of a Data Set (a viral Attribute has virality = TRUE, a non-viral Attribute has virality=FALSE, if the virality is not defined, the Attribute is considered as non-viral).

By default, an Attribute propagates from the operand Data Sets (DS_i) to the result Data Set (DS_r) if it is “viral” at least in one of the operand Data Sets. By default, an Attribute which is viral in one of the operands DS_i is considered as viral also in the result DS_r.
The Attribute propagation rule does not apply for the time series operators.

The Attribute propagation rule does not apply if the operations on the Attributes to be propagated are explicitly specified in the expression (for example through the `keep` and `calc` operators). This way it is possible to keep in the result also Attribute which are non-viral in all the operands, to drop viral Attributes, to override the (possible) default calculation algorithm of the Attribute, to change the virality of the resulting Attributes.
Parentheses: ( )

Syntax

(op)

Input parameters

- **op**: the operand to be evaluated before performing other operations written outside the parentheses.
- According to the general VTL rule, operators can be nested, therefore any Data Set, Component or scalar can be obtained through an expression as complex as needed (for example op can be written as the expression 2 + 3).

Examples of valid syntaxes

- (DS_1 + DS_2)
- (CMP_1 - CMP_2)
- (2 + DS_1)
- (DS_2 - 3 * DS_3)

Semantic for scalar operations

Parentheses override the default evaluation order of the operators that are described in the section "VTL-ML – Evaluation order of the Operators". The operations enclosed in the parentheses are evaluated first. For example (2+3)^4 returns 20, instead 2+3^4 returns 14 because the multiplication has higher precedence than the addition.

Input parameters type

<table>
<thead>
<tr>
<th>op ::</th>
<th>dataset</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>component</td>
</tr>
</tbody>
</table>

| | scalar |

Result type

<table>
<thead>
<tr>
<th>result ::</th>
<th>dataset</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>component</td>
</tr>
</tbody>
</table>

| | scalar |

Additional constraints

None.

Behaviour

As mentioned, the op of the parentheses can be obtained through an expression as complex as needed (for example op can be written as DS_1 - DS_2. The part of the expression inside the parentheses is evaluated before the part outside of the parentheses. If more parentheses are nested, the inner parentheses are evaluated first, for example (20 - 10 / (2 + 3)) * 3 would give 54.

Examples

- (DS_1 + DS_2) * DS_3
- (CMP_1 - CMP_2 / (CMP_3 + CMP_4)) * CMP_5

Persistent assignment: <-

Syntax

re <- op
Input Parameters

- **re**: the result
- **op**: the operand. According to the general VTL rule allowing the indentation of the operators, op can be obtained through an expression as complex as needed (for example, op can be the expression DS_1 - DS_2).

Examples of valid syntaxes

- DS_r <- DS_1
- DS_r <- DS_1 - DS_2

Semantics for scalar operations

**Input parameters type**

- **re**: name
- **op**: dataset

**Result type**

empty

**Additional constraints**

The assignment cannot be used at Component level because the result of a Transformation cannot be a Data Set Component. When operations at Component level are invoked, the result is the Data Set which the output Components belongs to.

**Behaviour**

The input operand op is assigned to the persistent result re, which assumes the same value as op. As mentioned, the operand op can be obtained through an expression as complex as needed (for example, op can be the expression DS_1 - DS_2).

The result re is a persistent Data Set that has the same data structure as the Operand. For example in DS_r <- DS_1 the data structure of DS_r is the same as the one of DS_1.

If the Operand op is a scalar value, the result Data Set has no Components and contains only such a scalar value. For example, income <- 3 assigns the value 3 to the persistent Data Set named income.

**Examples**

Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>2013</td>
<td>Belgium</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>2013</td>
<td>Denmark</td>
<td>2</td>
<td>10</td>
</tr>
<tr>
<td>2013</td>
<td>France</td>
<td>3</td>
<td>12</td>
</tr>
<tr>
<td>2013</td>
<td>Spain</td>
<td>4</td>
<td>20</td>
</tr>
</tbody>
</table>

**Example 1:** DS_r <- DS_1 results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>2013</td>
<td>Belgium</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>2013</td>
<td>Denmark</td>
<td>2</td>
<td>10</td>
</tr>
<tr>
<td>2013</td>
<td>France</td>
<td>3</td>
<td>12</td>
</tr>
<tr>
<td>2013</td>
<td>Spain</td>
<td>4</td>
<td>20</td>
</tr>
</tbody>
</table>
Non-persistent assignment: 

**Syntax**

\[ re := op \]

**Input parameters**

- `re`: the result
- `op`: the operand (according to the general VTL rule allowing the indentation of the operators, `op` can be obtained through an expression as complex as needed (for example `op` can be the expression `DS_1 - DS_2`).

**Examples of valid syntaxes**

- `DS_r := DS_1`
- `DS_r := 3`
- `DS_r := DS_1 - DS_2`
- `DS_r := 3 + 2`

**Semantic for scalar operations**

**Input parameters type**

- `re :: name`
- `op :: dataset | scalar`

**Result type**

- `empty`

**Additional constraints**

- The assignment cannot be used at Component level because the result of a Transformation cannot be a Data Set Component. When operations at Component level are invoked, the result is the Data Set which the output Components belongs to.
- The same symbol denoting the non-persistent assignment Operator (`:=`) is also used inside other operations at Component level (for example in `calc` and `aggr`) in order to assign the result of the operation to the output Component: please note that in these cases the symbol `:=` does not denote the non-persistent assignment (i.e., this Operator), which cannot operate at Component level, but a special keyword of the syntax of the other Operator in which it is used.

**Behaviour**

- The value of the operand `op` is assigned to the result `re`, which is non-persistent and therefore is not stored. As mentioned, the operand `op` can be obtained through an expression as complex as needed (for example `op` can be the expression `DS_1 - DS_2`).
- The result `re` is a non-persistent Data Set that has the same data structure as the Operand. For example in `DS_r := DS_1` the data structure of `DS_r` is the same as the one of `DS_1`.
- If the Operand `op` is a scalar value, the result Data Set has no Components and contains only such a scalar value. For example, `income := 3` assigns the value `3` to the non-persistent Data Set named `income`.

**Examples**

Given the operand Data Sets `DS_1`:

<table>
<thead>
<tr>
<th><code>Id_1</code></th>
<th><code>Id_2</code></th>
<th><code>Me_1</code></th>
<th><code>Me_2</code></th>
</tr>
</thead>
<tbody>
<tr>
<td>2013</td>
<td>Belgium</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>2013</td>
<td>Denmark</td>
<td>2</td>
<td>10</td>
</tr>
<tr>
<td>2013</td>
<td>France</td>
<td>3</td>
<td>12</td>
</tr>
<tr>
<td>2013</td>
<td>Spain</td>
<td>4</td>
<td>20</td>
</tr>
</tbody>
</table>
Example 1:

\[ DS_r := DS_1 \]

results in:

\[
\begin{array}{|c|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 & \text{Me}_2 \\
\hline
2013 & Belgium & 5 & 5 \\
2013 & Denmark & 2 & 10 \\
2013 & France & 3 & 12 \\
2013 & Spain & 4 & 20 \\
\hline
\end{array}
\]

Membership: \(#\)

Syntax

\[ ds\#\text{comp} \]

Input Parameters

- \(ds\): the Data Set
- \(\text{comp}\): the Data Set Component

Examples of valid syntaxes

- \(DS_1\#\text{COMP}_3\)

Semantic for scalar operations

This operator cannot be applied to scalar values.

Input parameters type

- \(ds\): dataset
- \(\text{comp}\): name < component >

Result type

- \(\text{result}\): dataset

Additional constraints

- \(\text{comp}\) must be a Data Set Component of the Data Set \(ds\)

Behaviour

The membership operator returns a Data Set having the same Identifier Components of \(ds\) and a single Measure. If \(\text{comp}\) is a Measure in \(ds\), then \(\text{comp}\) is maintained in the result while all other Measures are dropped.

If \(\text{comp}\) is an Identifier or an Attribute Component in \(ds\), then all the existing Measures of \(ds\) are dropped in the result and a new Measure is added. The Data Points’ values for the new Measure are the same as the values of \(\text{comp}\) in \(ds\). A default conventional name is assigned to the new Measure depending on its type: for example \text{num\_var} if the Measure is \text{numeric}, \text{string\_var} if it is \text{string} and so on (the default name can be renamed through the \text{rename} operator if needed).

The Attributes follow the Attribute propagation rule as usual (viral Attributes of \(ds\) are maintained in the result as viral, non-viral ones are dropped). If \(\text{comp}\) is an Attribute, it follows the Attribute propagation rule too.

The same symbol denoting the membership operator (\(#)\) is also used inside other operations at Component level (for example in \text{join, calc, aggr}) in order to identify the Components to be operated: please note that in these cases the symbol \(#)\) does not denote the membership operator (i.e., this operator, which does not operate at Component level), but a special keyword of the syntax of the other operator in which it is used.

Examples

Given the operand Data Set \(DS_1\):

1. Example 1:

\[ DS_r := DS_1 \]
<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>1</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>2</td>
<td>10</td>
<td>P</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>3</td>
<td>12</td>
<td></td>
</tr>
</tbody>
</table>

**Example 1:** \( DS_r := DS_1#Me_1 \) results in:

(assuming that At_1 is not viral in DS_1)

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>1</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>2</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>3</td>
</tr>
</tbody>
</table>

(assuming that At_1 is viral in DS_1)

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>2</td>
<td>P</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>3</td>
<td></td>
</tr>
</tbody>
</table>

**Example 2:** \( DS_r := DS_1#Id_1 \) assuming that At_1 is viral in DS_1 results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>num_var</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>1</td>
<td>P</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>2</td>
<td></td>
</tr>
</tbody>
</table>

**Example 3:** \( DS_r := DS_1#At_1 \) assuming that At_1 is viral in DS_1 results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>string_var</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td></td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>P</td>
<td>P</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**User-defined operator call**

**Syntax**

\[
\text{operatorName} ( \{ \text{argument} (, \text{argument})* \})
\]
Input parameters

1808 operatorName  the name of an existing user-defined operator
1809 argument  argument passed to the operator

Examples of valid syntaxes

1812 max1 ( 2, 3 )

Semantic for scalar operations

1815 It depends on the specific user-defined operator that is invoked.

Input parameters type

1818 operatorName :: name
1819 argument :: A data type compatible with the type of the parameter of the user-defined operator that
1820 is invoked (see also the “Type syntax” section).

Result type

1824 result :: The data type of the result of the user-defined operator that is invoked (see also the
1825 “Type syntax” section).

Additional constraints

1828 - operatorName must refer to an operator created with the define operator statement.
1829 - The type of each argument value must be compliant with the type of the corresponding parameter of the
1830 user defined operator (the correspondence is in the positional order).

Behaviour

1833 The invoked user-defined operator is evaluated. The arguments passed to the operator in the invocation are
1834 associated to the corresponding parameters in positional order, the first argument as the value of the first
1835 parameter, the second argument as the value of the second parameter, and so on. An underscore (“_”) can be
1836 used to denote that the value for an optional operand is omitted. One or more optional operands in the last
1837 positions can be simply omitted.

Examples

1840 Example 1:

1842 Definition of the max1 operator (see also “define operator” in the VTL-DL):

1844 define operator max1 ( x integer, y integer )
1845  returns boolean
1846  is if x > y then x else y
1847  end define operator

1849 User-defined operator call of the max1 operator:

1851 max1 ( 2, 3 )

Evaluation of an external routine : eval

Syntax

1856 eval ( externalRoutineName ( { argument } { , argument }* ) language languageName returns outputType )

Input parameters

1859 externalRoutineName  the name of an external routine
1860 argument  the arguments passed to the external routine
1861 language  the implementation language of the routine
1862 outputType  the data type of the object returned by eval (see outputParameterType in Data
1863 type syntax)
**Examples of valid syntaxes**

```plaintext
eval ( routine1 ( "eabcdefgh" ) language "PL/SQL" returns string )
```

**Semantics for scalar operations:**

This is not a scalar operation.

**Input parameters type**

- `externalRoutineName ::` name
- `argument ::` any data type
- `language ::` string
- `outputType ::` any data type restricting Data Set or scalar

**Result Type**

- `result ::` dataset

**Additional constraints**

- The `eval` is the only VTL Operator that does not allow nesting and therefore a Transformation can contain just one invocation of `eval` and no other invocations. In other words, `eval` cannot be nested as the operand of another operation as well as another operator cannot be nested as an operand of `eval`.
- The result of an expression containing `eval` must be persistent.
- `externalRoutineName` is the conventional name of a non-VTL routine.
- the invoked external routine must be consistent with the VTL principles, first of all its behaviour must be functional, so having in input and providing in output first-order functions.
- `argument` is an argument passed to the external routine, it can be a name or a value of a VTL artefacts or some other parameter required by the routine.
- the arguments passed to the routine correspond to the parameters of the invoked external routine in positional order; as usual the optional parameters are substituted by the underscore if missing. The conversion of the VTL input/output data types from and to the external routine processor is left to the implementation.

**Behaviour**

The `eval` operator invokes an external, non-VTL routine, and returns its result as a Data Set or a scalar. The specific data type can be given in the invocation. The routine specified in the `eval` operator can perform any internal logic.

**Examples**

Assuming that SQL3 is an SQL statement which produces `DS_r` starting from `DS_1`:

```plaintext
DS_r := eval( SQL3( DS_1 ) language "PL/SQL"
    returns dataset { identifier<geo_area> ref_area,
        identifier<date> time,
        measure<number> obs_value,
        attribute<string> obs_status } )
```

Assuming that f is an externally defined Java method:

```plaintext
DS_r := DS_1 [calc Me := eval ( f ( Me ) language "Java" returns integer) ]
```

**Type conversion:** 

**cast**

**Syntax**

```plaintext
cast ( op , scalarType ( , mask ) )
```

**Input parameters**

- `op` the operand to be cast
- `scalarType` the name of the scalar type into which `op` has to be converted
- `mask` a character literal that specifies the format of `op`
Examples of valid syntaxes
See the examples below.

Semantics for scalar operations:
This operator converts the scalar type of op to the scalar type specified by scalarType. It returns a copy of op converted to the specified scalarType.

Input parameters type
| op :: dataset{ measure<scalar> _ } |
| [ component<scalar> |
| scalar |
| scalarType :: scalar type |
| mask :: string |

Result type
| result :: dataset{ measure<scalar> _ } |
| [ component<scalar> |
| scalar |

Additional constraints
- Not all the conversions are possible, the specified casting operation is allowed only according to the semantics described below.
- The mask must adhere to one of the formats specified below.

Behaviour
Conversions between basic scalar types
The VTL assumes that a basic scalar type has a unique internal and more possible external representations (formats).

The external representations are those of the Value Domains which refers to such a basic scalar types (more Value Domains can refer to the same basic scalar type, see the VTL Data Types in the User Manual). For example, there can exist a boolean Value Domain which uses the values TRUE and FALSE and another boolean Value Domain which uses the values 1 and 0. The external representations are the ones of the Data Point Values and are obviously known by users.

The unique internal representation of a basic scalar type, instead, is used by the cast operator as a technical expedient to make the conversion between external representations easier: not necessarily users are aware of it.

In a conversion, the cast converts the source external representation into the internal representation (of the corresponding scalar type), then this last one is converted into the target external representation (of the target type). As mentioned in the User Manual, VTL does not prescribe any specific internal representation for the various scalar types, leaving different organisations free of using their preferred or already existing ones.

In some cases, depending on the type of op, the output scalarType and the invoked operator, an automatic conversion is made, that is, even without the explicit invocation of the cast operator: this kind of conversion is called implicit casting.

In other cases, more than all when the implicit casting is not possible, the type conversion must be specified explicitly through the invocation of the cast operator: this kind of conversion is called explicit casting. If an explicit casting is specified, the (possible) implicit casting is overridden. There are two main categories of explicit casting:
- "Explicit with mask": the explicit conversion uses a formatting mask that specifies how the actual casting is performed;
- "Explicit w/o mask": the explicit conversion does not use a formatting mask.

The table below summarises the possible castings between the basic scalar types. In particular, the input type is specified in the first column (row headings) and the output type in the first row (column headings).
The type of casting can be personalised in specific environments, provided that the personalisation is explicitly documented with reference to the table above. For example, assuming that an explicit cast with mask is required and that in a specific environment a definite mask is used for such a kind of conversions, the cast can also become implicit provided that the mask that will be applied is specified.

The implicit casting is performed when a value of a certain type is provided when another type is expected. Its behaviour is described here:

- From integer to number: an integer is provided when a number is expected (for example, an integer and a number are passed as inputs of a n-ary numeric operator); it returns a number having the integer part equal to the integer and the decimal part equal to zero;
- From integer to string: an integer is provided when a string is expected (for example, an integer is passed as an input of a string operator); it returns a string having the literal value of the integer;
- From number to string: a number is provided when a string is expected; it returns the string having the literal value of the number; the decimal separator is converted into the character "." (dot).
- From boolean to string: a boolean is provided when a string is expected; the boolean value TRUE is converted into the string "TRUE" and FALSE into the string "FALSE";
- From date to time: a date (point in time) is provided when a time is expected (interval of time): the conversion results in an interval having the same start and end, both equal to the original date;
- From time_period to time: a time_period (a regular interval of time, like a month, a quarter, a year ...) is provided when a time (any interval of time) is expected; it returns a time value having the same start and end as the time_period value.

An implicit cast is also performed from a value domain type or a set type to a basic scalar type: when a scalar value belonging to a Value Domains or a Set is involved in an operation (i.e., provided as input to an operator), the value is implicitly cast into the basic scalar type which the Value Domain refers to (for this relationship, see the description of Type System in the User Manual). For example, assuming that the Component birth_country is defined on the Value Domain country, which contains the ISO 3166-1 numeric codes and therefore refers to the basic scalar type integer, the (possible) invocation length(birth_country), which calculates the length of the input string, automatically casts the values of birth_country into the corresponding string. If the basic scalar type of the Value Domain is not compatible with the expression where it is used, an error is raised. This VTL feature is particularly important as it provides a general behaviour for the Value Domains and relevant Sets, preventing from the need of defining specific behaviours (or methods or operations) for each one of them. In other words, all the Values inherit the operations that can be performed on them from the basic scalar types of the respective Value Domains.

The cast operator can be invoked explicitly even for the conversions which allow an implicit cast and in this case the same behaviour as the implicit cast is applied.

The behaviour of the cast operator for the conversions that require explicit casting without mask is the following:

- From integer to boolean: if the integer is different from 0, then TRUE is returned, FALSE otherwise.

<table>
<thead>
<tr>
<th>Type</th>
<th>Implicit</th>
<th>Explicit w/o mask</th>
<th>Not feasible</th>
<th>Not feasible</th>
<th>Not feasible</th>
<th>Implicit</th>
<th>Not feasible</th>
</tr>
</thead>
<tbody>
<tr>
<td>integer</td>
<td>-</td>
<td></td>
<td>-</td>
<td>-</td>
<td>-</td>
<td></td>
<td></td>
</tr>
<tr>
<td>number</td>
<td>Explicit w/o mask</td>
<td>Explicit w/o mask</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Implicit</td>
<td>Not feasible</td>
</tr>
<tr>
<td>boolean</td>
<td>Explicit w/o mask</td>
<td>Explicit w/o mask</td>
<td>-</td>
<td>-</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
</tr>
<tr>
<td>time</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>-</td>
<td>Not feasible</td>
<td>Explicit with mask</td>
<td>Not feasible</td>
</tr>
<tr>
<td>date</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Implicit</td>
<td>-</td>
<td>Explicit with mask</td>
<td>Not feasible</td>
</tr>
<tr>
<td>time_period</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Implicit</td>
<td>Explicit with mask</td>
<td>-</td>
<td>Explicit with mask</td>
</tr>
<tr>
<td>string</td>
<td>Explicit w/o mask</td>
<td>Explicit with mask</td>
<td>Not feasible</td>
<td>Explicit with mask</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
</tr>
<tr>
<td>duration</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Not feasible</td>
<td>Explicit with mask</td>
<td>-</td>
</tr>
</tbody>
</table>
• From **number** to **integer**: converts a **number** with no decimal part into an **integer**; if the decimal part is present, a runtime error is raised.

• From **number** to **boolean**: if the **number** is different from 0.0, then TRUE is returned, FALSE otherwise.

• From **boolean** to **integer**: TRUE is converted into 1; FALSE into 0.

• From **boolean** to **number**: TRUE is converted into 1.0; FALSE into 0.0.

• From **date** to **time_period**: it converts a **date** into the corresponding daily value of **time_period**.

• From **string** to **integer**: the **integer** having the literal value of the **string** is returned; if the **string** contains a literal that cannot be matched to an **integer**, a runtime error is raised.

• From **string** to **time_period**: it converts a **string** value to a **time_period** value.

When an **explicit casting with mask** is required, the conversion is made by applying the formatting mask which specifies the meaning of the characters in the output **string**. The formatting Masks are described in the section “VTL-ML – Typical Behaviour of the ML Operators”, sub-section “Type Conversion and Formatting Mask.”

The behaviour of the **cast** operator for such conversions is the following:

• From **time** to **string**: it is applied the **time** formatting mask.

• From **time** to **string**: it is applied the **time_period** formatting mask.

• From **time_period** to **date**: it is applied a formatting mask which accepts two possible values (“START”, “END”). If “START” is specified, then the **date** is set to the beginning of the **time_period**; if “END” is specified, then the **date** is set to the end of the **time_period**.

• From **time_period** to **string**: it is applied the **time_period** formatting mask.

• From **duration** to **string**: a **duration** (an absolute time interval) is provided when a **string** is expected; it returns the **string** having the default **string** representation for the **duration**.

• From **string** to **number**: the **number** having the literal value of the **string** is returned; if the **string** contains a literal that cannot be matched to a **number**, a runtime error is raised. The **number** is generated by using a **number** formatting mask.

• From **string** to **time**: the **time** having the literal value of the **string** is returned; if the **string** contains a literal that cannot be matched to a **date**, a runtime error is raised. The **time** value is generated by using a **time** formatting mask.

• From **string** to **duration**: the **duration** having the literal value of the **string** is returned; if the **string** contains a literal that cannot be matched to a **duration**, a runtime error is raised. The **duration** value is generated by using a time formatting mask.

**Conversions between basic scalar types and Value Domains or Set types**

A value of a basic **scalar** type can be converted into a value belonging to a Value Domain which refers to such a **scalar** type. The resulting **scalar** value must be one of the allowed values of the Value Domain or Set; otherwise, a runtime error is raised. This specific use of **cast** operators does not really correspond to a type conversion; in more formal terms, we would say that it acts as a constructor, i.e., it builds an instance of the output type. Yet, towards a homogeneous and possibly simple definition of VTL syntax, we blur the distinction between constructors and type conversions and opt for a unique formalism. An example is given below.

**Conversions between different Value Domain types**

As a result of the above definitions, conversions between values of different Value Domains are also possible. Since an element of a Value Domain is implicitly cast into its corresponding basic **scalar** type, we can build on it to turn the so obtained **scalar** type into another Value Domain type. Of course, this latter Value Domain type must use as a base type this **scalar** type.

**Examples**

Example 1: from **string** to **number**

```
    ds2 := ds1[calc m2 := cast(m1, number, "DD.DDD") + 2 ]
```

In this case we use explicit cast from **string** to **numbers**. The mask is used to specify how the **string** must be interpreted in the conversion.

Example 2: from **string** to **date**

```
    ds2 := ds1[calc m2 := cast(m1, "YYYY-MM-DD") ]
```
In this case we use explicit cast from string to date. The mask is used to specify how the string must be interpreted in the conversion.

Example 3: from number to integer

\[
\text{ds2 := ds1[calc m2 := cast(m1, integer) + 3 ]}
\]

In this case we cast a number into an integer, no mask is required.

Example 4: from number to string

\[
\text{ds2 := ds1[calc m2 := length(cast(m1, string)) ]}
\]

In this case we cast a number into a string, no mask is required.

Example 5: from date to string

\[
\text{ds2 := ds1[calc m2 := cast(m1, string, "YY-MON-DAY hh:mm:ss") ]}
\]

In this example a date instant is turned into a string. The mask is used to specify the string layout.

Example 6: from string to GEO_AREA

\[
\text{ds2 := ds1[calc m2 := cast(GEO_STRING, GEO_AREA)]}
\]

In this example we suppose we have elements of Value Domain Subset for GEO_AREA. Let GEO_STRING be a string Component of Data Set ds1 with string values compatible with the GEO_AREA Value Domain Subset. Thus, the following expression moves ds1 data into ds2, explicitly casting strings to geographical areas.

Example 7: from GEO_AREA to string

\[
\text{ds2 := ds1[calc m2 := length(GEO_AREA)]}
\]

In this example we use a Component GEO_AREA in a string expression, which calculates the length of the corresponding string; this triggers the automatic cast.

Example 8: from GEO_AREA2 to GEO_AREA1

\[
\text{ds2 := ds1[ calc m2 := cast (GEO, GEO_AREA1) ]}
\]

In this example we suppose we have to compare elements two Value Domain Subsets, They are both defined on top of Strings. The following cast expressions performs the conversion. Now, Component GEO is of type GEO_AREA2, then we specify it has to be cast into GEO_AREA1. As both work on strings (and the values are compatible), the conversion is feasible. In other words, the cast of an operand into GEO_AREA1 would expect a string. Then, as GEO is of type GEO_AREA2, defined on top of strings, it is implicitly cast to the respective string; this is compatible with what cast expects and it is then able to build a value of type GEO_AREA1.

Example 9: from string to time_period

In the following examples we convert from strings to time_periods, by using appropriate masks.

The first quarter of year 2000 can be expressed as follows (other examples are possible):

\[
\text{cast ("2000Q1", time_period, "YYYY\'QQ")}
\]
\[
\text{cast ("2000-Q1", time_period, "YYYY\'-QQ")}
\]
\[
\text{cast ("2000-1", time_period, "YYYY-Q")}
\]
\[
\text{cast ("Q1-2000", time_period, "QQ-YYYY")}
\]
\[
\text{cast ("2000Q01", time_period, "YYYY\'QQQ")}
\]

Examples of daily data:

\[
\text{cast ("2000M01D01", time_period, "YYYY\MM\DDD")}
\]
\[
\text{cast ("2000.01.01", time_period, "YYYY.MM.DD")}
\]
The Join operators are fundamental VTL operators. They are part of the core of the language and allow to obtain the behaviour of the majority of the other non-core operators, plus many additional behaviours that cannot be obtained through the other operators.

The Join operators are four, namely the inner_join, the left_join, the full_join and the cross_join. Because their syntax is similar, they are described together.

**Join:** inner_join, left_join, full_join, cross_join

**Syntax**
```
joinOperator ( ds { as alias } { , ds { as alias } }* { using usingComp { , usingComp }* }  
  { filter filterCondition }  
  { apply applyExpr }  
  | calc calcClause  
  | aggr aggrClause { groupingClause }  
  { keep comp { , comp }* | drop comp { , comp }* }  
  { rename compFrom to compTo { , compFrom to compTo }* }  
) 
```

**Input parameters**
- **joinOperator** the Join operator to be applied
- **ds** the Data Set operands (at least one must be present)
- **alias** optional aliases for the input Data Sets, valid only within the "join" operation to make it easier to refer to them. If omitted, the Data Set name must be used.
- **usingComp** component of the input Data Sets whose values have to match in the join (the using clause is allowed for the left_join only under certain constraints described below and is not allowed at all for the full_join and cross_join)
- **filterCondition** a condition (boolean expression) at component level, having only Components of the input Data Sets as operands, which is evaluated for each joined Data Point and filters them (when TRUE the joined Data Point is kept, otherwise it is not kept)
- **applyExpr** an expression, having the input Data Sets as operands, which is pairwise applied to all their homonym Measure Components and produces homonym Measure Components in the result; for example if both the Data Sets ds1 and ds2 have the numeric measures m1 and m2, the clause apply ds1 + ds2 would result in calculating $m1 := ds1#m1 + ds2#m1$ and $m2 := ds1#m2 + ds2#m2$
- **calcClause** clause that specifies the Components to be calculated, their roles and their calculation algorithms, to be applied on the joined and filtered Data Points.
calcExpr
expression at component level, having only Components of the input Data Sets as
operands, used to calculate a Component

aggrClause
clause that specifies the required aggregations, i.e., the aggregated Components to be
calculated, their roles and their calculation algorithm, to be applied on the joined and
filtered Data Points

aggrRole
the role of the aggregated Component to be calculated; if omitted, the Measure role is
assumed

aggrComp
the name of the aggregated Component to be calculated; this is a dependent Component
of the result (Measure or Attribute, not Identifier)

aggrExpr
expression at component level, having only Components of the input Data Sets as
operands, which invokes an aggregate operator (e.g. \texttt{avg}, \texttt{count}, \texttt{max} ... , see also the
corresponding sections) to perform the desired aggregation. Note that the \texttt{count}
operator is used in an \texttt{aggrClause} without parameters, e.g.:

\begin{verbatim}
DS_1 [ aggr Me_1 := count () group by Id_1 ) ]
\end{verbatim}

groupingClause
the following alternative grouping options:

\begin{verbatim}
group by
\end{verbatim}
the Data Points are grouped by the values of the specified Identifiers
\begin{verbatim}
group except
\end{verbatim}
the Data Points are grouped by the values of the Identifiers not
specified as groupingId. The specified Identifiers are dropped in the
result.
\begin{verbatim}
group all
\end{verbatim}
converts the values of an Identifier Component using conversionExpr

and keeps all the resulting Identifiers.

groupingId
Identifier Component to be kept (in the \texttt{group by clause}) or dropped (in the \texttt{group}
except clause).

conversionExpr
specifies a conversion operator (e.g. \texttt{time_agg}) to convert an Identifier from finer to
coarser granularity. The conversion operator is applied on an Identifier of the operand
Data Set.

havingCondition
a condition (\texttt{boolean} expression) at component level, having only Components of the
input Data Sets as operands (and possibly constants), to be fulfilled by the groups of
Data Points: only groups for which havingCondition evaluates to \texttt{TRUE} appear in the
result. The havingCondition refers to the groups specified through the groupingClause,
therefore it must invoke aggregate operators (e.g. \texttt{avg}, \texttt{count}, \texttt{max}, ... , see also the
section Aggregate invocation). A correct example of havingCondition is

\begin{verbatim}
max(obs_value) < 1000,
\end{verbatim}

while the condition \texttt{obs_value < 1000} is not a right
havingCondition, because it refers to the values of single Data Points and not to the
groups. The count operator is used in a havingCondition without parameters, e.g.:

\begin{verbatim}
sum ( ds group by id1 having count () >= 10 )
\end{verbatim}

comp
dependent Component (Measure or Attribute, not Identifier) to be kept (in the \texttt{keep}
clause) or dropped (in the \texttt{drop} clause)

compFrom
the original name of the Component to be renamed

compTo
the new name of the Component after the renaming

\textit{Examples of valid syntaxes}

\begin{verbatim}
inner_join ( ds1 as d1, ds2 as d2 using ld1, ld2
filter d1#Me1 + d2#Me1 <10
apply d1 / d2
keep Me1, Me2, Me3
rename ld1 to ld10, id2 to id20
)
\end{verbatim}

\begin{verbatim}
left_join ( ds1 as d1, ds2 as d2
filter d1#Me1 + d2#Me1 <10
calc Me1 := d1#Me1 + d2#Me3
keep Me1
rename ld1 to Ident1, Me1 to Meas1
)
\end{verbatim}

\begin{verbatim}
full_join ( ds1 as d1, ds2 as d2
filter d1#Me1 + d2#Me1 <10
\end{verbatim}
aggr Me1 := sum(Me1), attribute At20 := avg(Me2)
group by  ld1, ld2
having sum(Me3) > 0 )

**Semantics for scalar operations**

The join operator does not perform scalar operations.

**Input parameters type**

ds:: dataset
alias :: name
usingld :: name < component >
filterCondition :: component<boolean>
applyExpr :: dataset
calcComp :: name < component >
calcExpr :: component<scalar>
aggrComp :: name < component >
aggrExpr :: component<scalar>
groupingld :: name < identifier >
conversionExpr :: component<scalar>
havingCondition :: component<boolean>
comp :: name < component >
compFrom :: component<scalar>
compTo :: component<scalar>

**Result type**

result :: dataset

**Additional constraints**

The aliases must be all distinct and different from the Data Set names. Aliases are mandatory for Data Sets which appear more than once in the Join (self-join) and for non-named Data Set obtained as result of a sub-expression. The using clause is not allowed for the full_join and for the cross_join, because otherwise a non-functional result could be obtained. If the using clause is not specified (we will label this case as “Case A”), calling Id(ds_i) the set of Identifier Components of operand ds_i, the following group of constraints must hold:

- For inner_join, for each pair ds_i, ds_j, either Id(ds_i) ⊆ Id(ds_j) or Id(ds_j) ⊆ Id(ds_i). In simpler words, the Identifiers of one of the joined Data Sets must be a superset of the identifiers of all the other ones.
- For left_join and full_join, for each pair ds_i, ds_j,  Id(ds_i) = Id(ds_j). In simpler words, the joined Data Sets must have the same Identifiers.
- For cross-join (Cartesian product), no constraints are needed.

If the using clause is specified (we will label this case as “Case B”, allowed only for the inner_join and the left_join), all the join keys must appear as Components in all the input Data Sets. Moreover two sub-cases are allowed:

- Sub-case B1: the constraints of the Case A are respected and the join keys are a subset of the common Identifiers of the joined Data Sets;
- Sub-case B2:
  - In case of inner_join, one Data Set acts as the reference Data Set which the others are joined to; in case of left_join, this is the left-most Data Set (i.e., ds1);
  - All the input Data Sets, except the reference Data Set, have the same Identifiers [ld1, ..., ld_n];
  - The using clause specifies all and only the common Identifiers of the non-reference Data Sets [ld1, ..., ld_n].

The join operators must fulfill also other constraints:

- **apply, calc and aggr** clauses are mutually exclusive
- **keep and drop** clauses are mutually exclusive
- **comp** can be only dependent Components (Measures and Attributes, not Identifiers)
- An Identifier not included in the **group by** clause (if any) cannot be included in the **rename** clause

---

7 These constraints hold also for the full_join and the cross_join, which do not allow the using clause.
- An Identifier included in the **group except** clause (if any) cannot be included in the **rename** clause. If the **aggr** clause is invoked and the grouping clause is omitted, no Identifier can be included in the **rename** clause.

- A dependent Component not included in the **keep** clause (if any) cannot be renamed.

- A dependent Component included in the **drop** clause (if any) cannot be renamed.

**Behaviour**

The semantics of the join operators can be procedurally described as follows.

- A relational join of the input operands is performed, according to SQL inner (**inner_join**), left-outer (**left_join**), full-outer (**full_join**) and Cartesian product (**cross_join**) semantics (these semantics will be explained below), producing an intermediate internal result, that is a Data Set that we will call “virtual” (**VDS**).

- The **filterCondition**, if present, is applied on **VDS**$_1$, producing the Virtual Data Set **VDS**$_2$.

- The specified calculation algorithms (**apply, calc** or **aggr**), if present, are applied on **VDS**$_2$. For the Attributes that have not been explicitly calculated in these clauses, the Attribute propagation rule is applied (see the User Manual), so producing the Virtual Data Set **VDS**$_3$.

- The **keep** or **drop** clause, if present, is applied on **VDS**$_3$, producing the Virtual Data Set **VDS**$_4$.

- The **rename** clause, if present, is applied on **VDS**$_4$, producing the Virtual Data Set **VDS**$_5$.

- The final automatic alias removal is performed in order to obtain the output Data Set.

An alias can be optionally declared for each input Data Set. The aliases are valid only within the “join” operation, in particular to allow joining a dataset with itself (self join). If omitted, the input Data Sets are referenced only through their Data Set names. If the aliases are ambiguous (for example duplicated or equal to the name of another Data Set), an error is raised.

The structure of the virtual Data Set **VDS**$_1$, which is the output of the relational join is the following.

For the **inner_join**, the **left_join** and the **full_join**, the virtual Data Set contains the following Components:

- The Components used as join keys, which appear once and maintain their original names and roles. In the cases A and B1, all of them are Identifiers. In the sub-case B2, the result takes the roles from the reference Data Set.

- In the sub-case B2: the Identifiers of the reference Data Set, which appear once and maintain their original name and role.

- The other Components coming from exactly one input Data Set, which appear once and maintain their original name.

- The other Components coming from more than one input Data Set, which appears as many times as the Data Set they come from; to distinguish them, their names are prefixed with the alias (or the name) of the Data Set they come from, separated by the “#” symbol (e.g., **ds#cmp**). For example, if the Component “population” appears in two input Data Sets “ds1” and “ds2” that have the aliases “a” and “b” respectively, the Components “a#population” and “b#population” will appear in the virtual Data Set.

- If the Components are not defined, the two Components are prefixed with the Data Set name (i.e., “ds1#population” and “ds2#population”). In this context, the symbol “#” does not denote the membership operator but acts just as a separator between the the Data Set and the Component names.

- If the same Data Set appears more times as operand of the join (self join) and the aliases are not defined, an exception is raised because it is not allowed that two or more Components in the virtual Data Set have the same name. In the self-join the aliases are mandatory to disambiguate the Component names.

- If a Data Set in the join list is the result of a sub-expression, then an alias is mandatory all the same because this Data Set has no name. If the alias is omitted, an exception is raised.

As for the **cross_join**, the virtual Data Set contains all the Components from all the operands, possibly prefixed with the aliases to avoid ambiguities.

The semantics of the relational join is the following.

The join is performed on some join keys, which are the Components of the input Data Sets whose values are used to match the input Data Points and produce the joined output Data Points.

By default (only for the **full_join** and the **cross_join**), the join is performed on the subset of homonym Identifier Components of the input Data Sets.

The parameter **using** allows to specify different join keys than the default ones, and can be used only for the **inner_join** and the **left_join** in order to preserve the functional behaviour of the operations.

The different kinds of relational joins behave as follows.

- **inner_join**: the Data Points of **ds1, ..., dsN** are joined if they have the same values for the common Identifier Components or, if the **using** clause is present, for the specified Components. A (joined) virtual Data Point is generated in the virtual Data Set **VDS**$_1$ when a matching Data Point is found for each one of the input Data Sets. In this case, the Values of the Components of a virtual Data Point are taken from the
corresponding Components of the matching Data Points. If there is no match for one or more input Data Sets, no virtual Data Point is generated.

- **left_join**: the join is ideally performed stepwise, between consecutive pairs of input Data Sets, starting from the left side and proceeding towards the right side. The Data Points are matched like in the *inner_join*, but a virtual Data Point is generated even if no Data Point of the right Data Set matches (in this case, the Measures and Attributes coming from the right Data Set take the NULL value in the virtual Data Set). Therefore, for each Data Points of the left Data Set a virtual Data Point is always generated. These stepwise operations are associative. More formally, consider the generic pair <ds<sub>i</sub>, ds<sub>i+1</sub>>, where ds<sub>i</sub> is the result of the left join of the first "i" operands and ds<sub>i+1</sub> is the i+1<sup>th</sup> operand. For each pair <ds<sub>i</sub>, ds<sub>i+1</sub>>, the joined Data Set is fed with all the Data Points that match in ds<sub>i</sub> and ds<sub>i+1</sub> or are only in ds<sub>i</sub>. The constraints described above guarantee the absence of null values for the Identifier Components of the joined Data Set, whose values are always taken from the left Data Set. If the join succeeds for a Data Point in ds<sub>i</sub>, the values for the Measures and the Attributes are carried from ds<sub>i</sub> and ds<sub>i+1</sub> as explained above. Otherwise, i.e., if no Data Point in ds<sub>i+1</sub> matches the Data Point in ds<sub>i</sub>, null values are given to Measures and Attributes coming only from ds<sub>i+1</sub>.

- **full_join**: the join is ideally performed stepwise, between consecutive pairs of input Data Sets, starting from the left side and proceeding toward the right side. The Data Points are matched like in the *inner_join* and *left_join*, but the *using* clause is not allowed and a virtual Data Point is generated either if no Data Point of the right Data Set matches with the left Data Point or if no Data Point of the left Data Set matches with the right Data Point (in this case, Measures and Attributes coming from the non matching Data Set take the NULL value in the virtual Data Set). Therefore, for each Data Points of the left and the right Data Set, a virtual Data Point is always generated. These stepwise operations are associative. More formally, consider the generic pair <ds<sub>i</sub>, ds<sub>i+1</sub>>, where ds<sub>i</sub> is the result of the *full_join* of the first "i" operands and ds<sub>i+1</sub> is the i+1<sup>th</sup> operand. For each pair <ds<sub>i</sub>, ds<sub>i+1</sub>>, the resulting Data Set is fed with the Data Points that match in ds<sub>i</sub> and ds<sub>i+1</sub> or that are only in ds<sub>i</sub> or in ds<sub>i+1</sub>. If for a Data Point in ds<sub>i</sub> the join succeeds, the values for the Measures and the Attributes are carried from ds<sub>i</sub> and ds<sub>i+1</sub> as explained. Otherwise, i.e., if no Data Point in ds<sub>i+1</sub> matches the Data Point in ds<sub>i</sub>, NULL values are given to Measures and Attributes coming only from ds<sub>i+1</sub>. Symmetrically, if no Data Point in ds<sub>i</sub> matches the Data Point in ds<sub>i+1</sub>, NULL values are given to Measures and Attributes coming only from ds<sub>i</sub>. The constraints described above guarantee the absence of NULL values on the Identifier Components. As mentioned, the *using* clause is not allowed in this case.

- **cross_join**: the join is performed stepwise, between consecutive pairs of input Data Sets, starting from the left side and proceeding toward the right side. No match is performed but the Cartesian product of the input Data Points is generated in output. These stepwise operations are associative. More formally, consider the ordered pair <ds<sub>i</sub>, ds<sub>i+1</sub>>, where ds<sub>i</sub> is the result of the *cross_join* of the first "i" operands and ds<sub>i+1</sub> is the i+1<sup>th</sup> operand. For each pair <ds<sub>i</sub>, ds<sub>i+1</sub>>, the resulting Data Set is fed with the Data Points obtained as the Cartesian product between the Data Points of ds<sub>i</sub> and ds<sub>i+1</sub>. The resulting Data Set will have all the Components from ds<sub>i</sub> and ds<sub>i+1</sub>. For the Data Sets which have at least one Component in common, the alias parameter is mandatory. As mentioned, the *using* parameter is not allowed in this case.

The semantics of the clauses is the following.

- **filter** takes as input a Boolean Component expression (having type *component<boolean>*). This clause filters in or out the input Data Points; when the expression is TRUE the Data Point is kept, otherwise it is not kept in the result. Only one filter clause is allowed.

- **apply** combines the homonym Measures in the source operands whose type is compatible with the operators used in *applyExpr*, generating homonym Measures in the output. The expression *applyExpr* can use as input the names or aliases of the operand Data Sets. It applies the expression to all the n-uples of homonym Measures in the input Data Sets producing in the target a single homonym Measure for each n-uple. It can be thought of as the multi-measure version of the calc. For example, if the following aliases have been declared: d1, d2, d3, then the following expression d1+d2+d3, sums all the homonym Measures in the three input Data Sets, say M1 and M2, so as to obtain in the result: M1 := d1#M1 + d2#M1 + d3#M1 and M2 := d1#M2 + d2#M2 + d3#M2. It is not only a compact version of a multiple calc, but also essential when the number of Measures in the input operands is not known beforehand. Only one apply clause is allowed.

- **calc** calculates new Identifier, Measure or Attribute Components on the basis of sub-expressions at Component level. Each Component is calculated through an independent sub-expression. It is possible to specify the role of the calculated Component among measure, identifier, attribute, or viral attribute, therefore the calc clause can be used also to change the role of a Component when possible. The keyword viral allows controlling the virality of Attributes (for the Attribute propagation rule see the User Manual). The following rule is used when the role is omitted: if the component exists in the operand Data Set then it maintains that role; if the component does not exist in the operand Data Set then the role is measure. The calcExpr are independent one another, they can only reference...
Components of the input Virtual Data Set and cannot use Components generated, for example, by other calcExpr. If the calculated Component is a new Component, it is added to the output virtual Data Set. If the calculated component is a Measure or an Attribute that already exists in the input virtual Data Set, the calculated values overwrite the original values. If the Calculated component is an Identifier that already exists in the input virtual Data Set, an exception is raised because overwriting an Identifier Component is forbidden for preserving the functional behaviour. Analytic operators can be used in the calc clause.

- **aggr** calculates aggregations of dependent Components (Measures or Attributes) on the basis of sub-expressions at Component level. Each Component is calculated through an independent sub-expression. It is possible to specify the role of the calculated Component among measure, identifier, attribute, or **viral** attribute. The substring viral allows to control the virality of Attributes, if the Attribute propagation rule is adopted (see the User Manual). The aggr sub-expressions are independent of one another, they can only reference Components of the input Virtual Data Set and cannot use Components generated, for example, by other aggr sub-expressions. The aggr computed Measures and Attributes are the only Measures and Attributes returned in the output virtual Data Set (plus the possible viral Attributes, see below Attribute propagation). The sub-expressions must contain only Aggregate operators, which are able to compute an aggregated Value relevant to a group of Data Points. The groups of Data Points to be aggregated are specified through the groupingClause, which allows the following alternative options.

  - **group by** the Data Points are grouped by the values of the specified Identifier. The Identifiers not specified are dropped in the result.
  - **group except** the Data Points are grouped by the values of the Identifiers not specified in the clause.
  - **group all** converts an Identifier Component using conversionExpr and keeps all the resulting Identifiers.

The having clause is used to filter groups in the result by means of an aggregate condition evaluated on the single groups, for example the minimum number of rows in the group. If no grouping clause is specified, then all the input Data Points are aggregated in a single group and the clause returns a Data Set that contains a single Data Point and has no Identifier Components.

- **keep** maintains in the output only the specified dependent Components (Measures and Attributes) of the input virtual Data Set and drops the non-specified ones. It has the role of a projection in the usual relational semantics (specifying which columns have to be projected in). Only one keep clause is allowed. If keep is used, drop must be omitted.

- **drop** maintains in the output only the non-specified dependent Components (Measures and Attributes) of the input virtual Data Set (component<scalar>) and drops the specified ones. It has the role of a projection in the usual relational join semantics (specifying which columns will be projected out). Only one drop clause is allowed. If drop is used, keep must be omitted.

- **rename** assigns new names to one or more Components (Identifier, Measure or Attribute Components).

The resulting Data Set, after renaming all the specified Components, must have unique names of all its Components (otherwise a runtime error is raised). Only the Component name is changed and not the Component Values, therefore the new Component must be defined on the same Value Domain and Value Domain Subset as the original Component (see also the IM in the User Manual). If the name of a Component defined on a different Value Domain or Set is assigned, an error is raised. In other words, rename is a transformation of the variable without any change in its values.

The semantics of the Attribute propagation in the join is the following. The Attributes calculated through the calc or aggr clauses are maintained unchanged. For all the other Attributes that are defined as viral, the Attribute propagation rule is applied (for the semantics, see the Attribute Propagation Rule section in the User Manual). This is done before the application of the drop, keep and rename clauses, which acts also on the Attributes resulting from the propagation.

The semantics of the final automatic aliases removal is the following. After the application of all the clauses, the structure of the final virtual Data Set is further modified. All the Components of the form "alias#component_name" (or "dataset_name#component_name") are implicitly renamed into "component_name". This means that the prefixes in the Component names are automatically removed. It is responsibility of the user to guarantee the absence of duplicated Component names once the prefixes are removed. In other words, the user must ensure that there are no pairs of Components whose names are of the form "alias1#c1" and "alias2#c1" in the structure of the virtual Data Point, since the removal of "alias1" and "alias2" would cause the clash. If, after the aliases removal two Components have the same name, an error is raised. In particular, name conflicts may derive if the using clause is present and some homonym Identifier Components do not appear in it; these components should be properly renamed because cannot be removed; the
input Data Set have homonym Measures and there is no apply clause which unifies them; these Measures can be renamed or removed.

**Examples**

Given the operand Data Sets DS_1 and DS_2:

<table>
<thead>
<tr>
<th>DS_1</th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
<td>Id_2</td>
<td>Me_1</td>
<td>Me_2</td>
</tr>
<tr>
<td>1</td>
<td>A</td>
<td>A</td>
<td>B</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>C</td>
<td>D</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>E</td>
<td>F</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>DS_2</th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
<td>Id_2</td>
<td>Me_1A</td>
<td>Me_2</td>
</tr>
<tr>
<td>1</td>
<td>A</td>
<td>B</td>
<td>Q</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>C</td>
<td>T</td>
</tr>
<tr>
<td>3</td>
<td>A</td>
<td>Z</td>
<td>M</td>
</tr>
</tbody>
</table>

**Example 1:**

DS_r := inner_join ( DS_1 as d1, DS_2 as d2
keep Me_1, d2#Me_2, Me_1A)  results in:

| DS_r | | | | |
|-------|-------|-------|-------|
| Id_1  | Id_2  | Me_1  | Me_2  | Me_1A |
| 1     | A     | A     | Q     | B     |
| 1     | B     | C     | T     | S     |

**Example 2:**

DS_r := left_join ( DS_1 as d1, DS_2 as d2
keep Me_1, d2#Me_2, Me_1A ) results in:

| DS_r | | | | |
|-------|-------|-------|-------|
| Id_1  | Id_2  | Me_1  | Me_2  | Me_1A |
| 1     | A     | A     | Q     | B     |
| 1     | B     | C     | T     | S     |
| 2     | A     | null  | null  | null  |

**Example 3:**

DS_r := full_join ( DS_1 as d1, DS_2 as d2
keep Me_1, d2#Me_2, Me_1A ) results in:

| DS_r | | | | |
|-------|-------|-------|-------|
| Id_1  | Id_2  | Me_1  | Me_2  | Me_1A |
| 1     | A     | A     | Q     | B     |
| 1     | B     | C     | T     | S     |
| 2     | A     | null  | null  | null  |
Example 4:

```
DS_r := cross_join (DS_1 as d1, DS_2 as d2
    rename d1#Id_1 to Id11, d1#Id_2 to Id12, d2#Id1 to Id21, d2#Id2 to Id22, d1#Me_2
to Me12 )
```

results in:

<table>
<thead>
<tr>
<th>Id_11</th>
<th>Id_12</th>
<th>Id_21</th>
<th>Id_22</th>
<th>Me_1</th>
<th>Me12</th>
<th>Me_1A</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>1</td>
<td>A</td>
<td>A</td>
<td>B</td>
<td>B</td>
<td>Q</td>
</tr>
<tr>
<td>1</td>
<td>A</td>
<td>1</td>
<td>B</td>
<td>A</td>
<td>B</td>
<td>S</td>
<td>T</td>
</tr>
<tr>
<td>1</td>
<td>A</td>
<td>3</td>
<td>A</td>
<td>A</td>
<td>B</td>
<td>Z</td>
<td>M</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>1</td>
<td>A</td>
<td>C</td>
<td>D</td>
<td>B</td>
<td>Q</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>1</td>
<td>B</td>
<td>C</td>
<td>D</td>
<td>S</td>
<td>T</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>3</td>
<td>A</td>
<td>C</td>
<td>D</td>
<td>Z</td>
<td>M</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>1</td>
<td>A</td>
<td>E</td>
<td>F</td>
<td>B</td>
<td>Q</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>1</td>
<td>B</td>
<td>E</td>
<td>F</td>
<td>S</td>
<td>T</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>3</td>
<td>A</td>
<td>E</td>
<td>F</td>
<td>Z</td>
<td>M</td>
</tr>
</tbody>
</table>

Example 5:

```
DS_r := inner_join (DS_1 as d1, DS_2 as d2
    filter Me_1 = "A"
    calc Me_4 = Me_1 || Me_1A
    drop d1#Me_2)
```

where || is the string concatenation, results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_1A</th>
<th>Me_4</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>B</td>
<td>A</td>
<td>Q</td>
<td>B</td>
<td>AB</td>
</tr>
</tbody>
</table>

Example 6:

```
DS_r := inner_join ( DS_1
calc Me_2 := Me_2 || "_NEW"
filter Id_2 ="B"
keep Me_1, Me_2)
```

where || is the string concatenation, results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>B</td>
<td>C</td>
<td>D_NEW</td>
</tr>
</tbody>
</table>

Example 7:

Given the operand Data Sets DS_1 and DS_2:
### DS_1

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>A</td>
<td>B</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>C</td>
<td>D</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>E</td>
<td>F</td>
</tr>
</tbody>
</table>

### DS_2

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>B</td>
<td>Q</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>S</td>
<td>T</td>
</tr>
<tr>
<td>3</td>
<td>A</td>
<td>Z</td>
<td>M</td>
</tr>
</tbody>
</table>

\[
DS_r := \text{inner_join ( DS_1 as d1, DS_2 as d2)} \\text{apply d1 || d2)}
\]

### DS_r

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>AB</td>
<td>BQ</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>CS</td>
<td>DT</td>
</tr>
</tbody>
</table>
String concatenation: `||`

**Syntax**

```
op1 || op2
```

**Input Parameters**

- `op1`, `op2`: the operands

**Examples of valid syntaxes**

- "Hello" || "world!"
- `ds_1 || ds_2`

**Semantics for scalar operations**

Concatenates two strings. For example, "Hello" || "world!" gives "Hello, world!"

**Input parameters type**

```
op1, op2 ::
  dataset { measure<string> _+ }
  | component<string>
  | string
```

**Result type**

```
result ::
  dataset { measure<string> _+ }
  | component<string>
  | string
```

**Additional constraints**

None.

**Behaviour**

The operator has the behaviour of the "Operators applicable on two Scalar Values or Data Sets or Data Set Components" (see the section “Typical behaviours of the ML Operators”).

**Examples**

Given the Data_Sets `DS_1` and `DS_2`:

**Table DS_1**

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;hi&quot;</td>
</tr>
</tbody>
</table>

**Table DS_2**

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;world&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;there&quot;</td>
</tr>
</tbody>
</table>

**Example 1**: `DS_r := DS_1 || DS_2` results in:
Example 2 (on component): $DS_r := DS_1[calc Me_2:= Me_1 \ || \ " \ " \ world"]$ results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;hi&quot;</td>
<td>&quot;hi world&quot;</td>
</tr>
</tbody>
</table>

Whitespace removal: trim, rtrim, ltrim

Syntax

```plaintext
{trim|rtrim|rtrim} ( op )
```

Input parameters

- `op`: the operand

Examples of valid syntaxes

- `trim("Hello ")`
- `trim(ds_1)`

Semantics for scalar operations

Removes trailing or/and leading whitespace from a string. For example, `trim("Hello ")` gives "Hello".

Input parameters type

```plaintext
op ::
| dataset { measure<string> _+ }
| component<string>
| string
```

Result type

```plaintext
result ::
| dataset { measure<string> _+ }
| component<string>
| string
```

Additional constraints

None.

Behaviour

The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

Examples

Given the Data Set $DS_1$:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello   &quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;hi    &quot;</td>
</tr>
</tbody>
</table>
Example 1: \( DS_r := \text{rtrim}(DS_1) \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;hi&quot;</td>
</tr>
</tbody>
</table>

Example 2 (on component): \( DS_r := DS_1[\text{calc}\ Me_2 := \text{rtrim}(Me_1)] \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
<td>&quot;hello&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;hi&quot;</td>
<td>&quot;hi&quot;</td>
</tr>
</tbody>
</table>

Character case conversion: upper/lower

Syntax

\[
\{\text{upper} | \text{lower}\} (\text{op})
\]

Input Parameters

- \text{op} the operand

Examples of valid syntaxes

- upper("Hello")
- lower(ds_1)

Semantics for scalar operations

Converts the character case of a string in upper or lower case. For example, upper("Hello") gives "HELLO".

Input Parameters type

- \text{op} :: dataset { measure<string> \_+ } |
  component<string> |
  string

Result type

- \text{result} :: dataset { measure<string> \_+ } |
  component<string> |
  string

Additional constraints

None.

Behaviour

The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

Examples

Given the Data Set \( DS_1 \):

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;hi&quot;</td>
</tr>
</tbody>
</table>
Example 1: \( DS_r := \text{upper}(DS_1) \) results in:

<table>
<thead>
<tr>
<th>( Id_1 )</th>
<th>( Id_2 )</th>
<th>( Me_1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;HELLO&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;HI&quot;</td>
</tr>
</tbody>
</table>

Example 2 (on component): \( DS_r := DS_1[\text{calc } Me_2 := \text{upper}(Me_1)] \) results in:

<table>
<thead>
<tr>
<th>( Id_1 )</th>
<th>( Id_2 )</th>
<th>( Me_1 )</th>
<th>( Me_2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
<td>&quot;HELLO&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>&quot;hi&quot;</td>
<td>&quot;HI&quot;</td>
</tr>
</tbody>
</table>

Sub-string extraction: \( \text{substr} \)

Syntax

\[
\text{substr} \ ( \text{op}, \text{start}, \text{length} )
\]

Input parameters

- \( \text{op} \): the operand
- \( \text{start} \): the starting digit (first character) of the string to be extracted
- \( \text{length} \): the length (number of characters) of the string to be extracted

Examples of valid syntaxes

- \( \text{substr} \ ( DS_1, 2 , 3 ) \)
- \( \text{substr} \ ( DS_1, 2 ) \)
- \( \text{substr} \ ( DS_1, _ , 3 ) \)
- \( \text{substr} \ ( DS_1 ) \)

Semantics for scalar operations

The operator extracts a substring from \( \text{op} \), which must be \text{string} type. The substring starts from the \( \text{start} \)th character of the input string and has a number of characters equal to the \( \text{length} \) parameter.

- If \( \text{start} \) is omitted, the substring starts from the 1st position.
- If \( \text{length} \) is omitted or overcomes the length of the input string, the substring ends at the end of the input string.
- If \( \text{start} \) is greater than the length of the input string, an empty string is extracted.

For example:

- \( \text{substr} \ ( \text{"abcdefghijklmnopqrstuvwxyz"}, 5 , 10 ) \) gives: "efghijklmn".
- \( \text{substr} \ ( \text{"abcdefghijklmnopqrstuvwxyz"}, 25 , 10 ) \) gives: "yz".
- \( \text{substr} \ ( \text{"abcdefghijklmnopqrstuvwxyz"}, 30 , 10 ) \) gives: "".

Input parameters type

- \( \text{op} :: \) dataset \{ measure <string> _+ \}
- \( \text{start} :: \) component <integer \[\text{value} >= 1\] >

length :: component < integer [ value >= 0 ] >
          | integer [ value >= 0 ]

Result type
result :: dataset { measure<string> _+ }
          | component<string>
          | string

Additional constraints
None.

Behaviour
As for the invocations at Data Set level, the operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component”, as for the invocations at Component or Scalar level, the operator has the behaviour of the “Operators applicable on more than two Scalar Values or Data Set Components”, (see the section “Typical behaviours of the ML Operators”).

Examples

Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello world&quot;</td>
<td>&quot;medium size text&quot;</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>&quot;abcdefghilmno&quot;</td>
<td>&quot;short text&quot;</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>&quot;pqrsuvwxyz&quot;</td>
<td>&quot;this is a long description&quot;</td>
</tr>
</tbody>
</table>

Example 1: DS_r:= substr ( DS_1, 7 ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
</table>
| 1    | A    | "world" | " size text"
| 1    | B    | "ghilmno" | "text"      |
| 2    | A    | "vwxyz" | "s a long description" |

Example 2: DS_r:= substr ( DS_1, 1, 5 ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
</table>
| 1    | A    | "hello" | "mediu"
| 1    | B    | "abcde" | "short" |
| 2    | A    | "pqrst" | "this " |

Example3(on Components): DS_r:= DS_1 [ calc Me_2:= substr ( Me_2, 1, 5 ) ] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello world&quot;</td>
<td>&quot;mediu&quot;</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>&quot;abcdefghilmno&quot;</td>
<td>&quot;short&quot;</td>
</tr>
</tbody>
</table>
String pattern replacement: replace

Syntax

\[ \text{replace} \left( \text{op}, \text{pattern1}, \text{pattern2} \right) \]

Input parameters

- \text{op} the operand
- \text{pattern1} the pattern to be replaced
- \text{pattern2} the replacing pattern

Examples of valid syntaxes

- replace(DS_1, "Hello", "Hi")
- replace(DS_1, "Hello")

Semantics for scalar operations

Replaces all the occurrences of a specified string-pattern (\text{pattern1}) with another one (\text{pattern2}). If \text{pattern2} is omitted then all occurrences of \text{pattern1} are removed. For example:

- replace("Hello world", "Hello", "Hi") gives "Hi world"
- replace("Hello world", "Hello") gives " world"
- replace("Hello", "ello", "i") gives "Hi"

Input parameters type

- \text{op} :: dataset \{ measure<string> \_+ \}
  \hspace{1cm} | \hspace{1cm} component<string>
  \hspace{1cm} | \hspace{1cm} string
- \text{pattern1, pattern2} :: component<string>
  \hspace{1cm} | \hspace{1cm} string

Result type

- \text{result} :: dataset \{ measure<string> \_+ \}
  \hspace{1cm} | \hspace{1cm} component<string>
  \hspace{1cm} | \hspace{1cm} string

Additional constraints

None.

Behaviour

As for the invocations at Data Set level, the operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component”, as for the invocations at Component or Scalar level, the operator has the behaviour of the “Operators applicable on more than two Scalar Values or Data Set Components”, (see the section “Typical behaviours of the ML Operators”).

Examples

given the Data_Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>3</td>
<td>A</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>4</td>
<td>A</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>
Example 1: \( DS_r := \text{replace}(ds_1,"ello","i") \) results in:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hi world&quot;</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>&quot;say hi&quot;</td>
</tr>
<tr>
<td>3</td>
<td>A</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>4</td>
<td>A</td>
<td>&quot;hi! &quot;</td>
</tr>
</tbody>
</table>

Example 2 (on component): \( DS_r := DS_1[ \text{calc Me}_2 := \text{replace}(Me_1,"ello","i") ] \) results in:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello world&quot;</td>
<td>&quot;hi world&quot;</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>&quot;say hello&quot;</td>
<td>&quot;say hi&quot;</td>
</tr>
<tr>
<td>3</td>
<td>A</td>
<td>&quot;he&quot;</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>4</td>
<td>A</td>
<td>&quot;hello! &quot;</td>
<td>&quot;hi! &quot;</td>
</tr>
</tbody>
</table>

String pattern location: \( \text{instr} \)

**Syntax**

\[ \text{instr} \left( \text{op}, \text{pattern}, \text{start}, \text{occurrence} \right) \]

**Input parameters**

- \text{op}: the operand
- \text{pattern}: the string-pattern to be searched
- \text{start}: the position in the input string of the character from which the search starts
- \text{occurrence}: the occurrence of the pattern to search

**Examples of valid syntaxes**

- \text{instr}( DS_1, "ab", 2, 3 )
- \text{instr}( DS_1, "ab", 2 )
- \text{instr}( DS_1, "ab", _, 2 )
- \text{instr}( DS_1, "ab" )

**Semantics for scalar operations**

The operator returns the position in the input string of a specified string (pattern). The search starts from the start\(\text{th}\) character of the input string and finds the \text{n}\text{th} occurrence of the pattern, returning the position of its first character.

- If \text{start} is omitted, the search starts from the 1\text{st} position.
- If \text{n}\text{th} occurrence is omitted, the value is 1.
- If the \text{n}\text{th} occurrence of the string-pattern after the start\text{th} character is not found in the input string, the returned value is 0.

For example:

- \text{instr}("abcde","c") gives 3
- \text{instr}("abcdefxcwds","c", _, 3 ) gives 10
- \text{instr}("abcdefcxcwds","c", 5, 3 ) gives 0

**Input parameters type**
op :: dataset { measure<string> _ }  
| component<string>  
| string  

pattern :: component<string>  
| string  

start :: component < integer [ value >= 1 ] >  
| integer [ value >= 1 ]  

occurrence :: component < integer [ value >= 1 ] >  
| integer [ value >= 1 ]  

Result type  
result :: dataset { measure<integer[value >= 0]> int_var }  
| component<integer[value >= 0]>  
| integer[value >= 0]  

Additional constraints  
For operations at Data Set level, the input Data Set must have exactly one string type Measure.  

Behaviour  
As for the invocations at Data Set level, the operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component”, as for the invocations at Component or Scalar level, the operator has the behaviour of the “Operators applicable on more than two Scalar Values or Data Set Components”, (see the section “Typical behaviours of the ML Operators”).  

If op is a Data Set then instr returns a dataset with a single measure int_var of type integer.  

Examples  
Given the Data Set DS_1:  

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>3</td>
<td>A</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>4</td>
<td>A</td>
<td>&quot;hi, hello! &quot;</td>
</tr>
</tbody>
</table>

Example 1:  DS_r := instr(ds_1,"hello")  results in  

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>int_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>5</td>
</tr>
<tr>
<td>3</td>
<td>A</td>
<td>0</td>
</tr>
<tr>
<td>4</td>
<td>A</td>
<td>5</td>
</tr>
</tbody>
</table>

Example 2 (on component):  DS_r := DS_1[calc Me_2:=instr(Me_1,"hello")]]  results in:  

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello world&quot;</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>&quot;say hello&quot;</td>
<td>5</td>
</tr>
<tr>
<td>3</td>
<td>A</td>
<td>&quot;he&quot;</td>
<td>0</td>
</tr>
<tr>
<td>4</td>
<td>A</td>
<td>&quot;hi, hello! &quot;</td>
<td>5</td>
</tr>
</tbody>
</table>
Given the Data Set DS_2:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
<td>&quot;world&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>NULL</td>
<td>&quot;hi&quot;</td>
</tr>
</tbody>
</table>

**Example 3 (applying the instr operator at component level to a multi Measure Data Set):**

\[ DS_r := DS_2 \left[ \text{calc } Me_10 := \text{instr}(Me_1, "o"), Me_20 := \text{instr}(Me_2, "o") \right] \]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_10</th>
<th>Me_20</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
<td>&quot;world&quot;</td>
<td>5</td>
<td>2</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>NULL</td>
<td>&quot;hi&quot;</td>
<td>null</td>
<td>0</td>
</tr>
</tbody>
</table>

**Example 4 (applying the instr operator at Data Set level to a multi Measure Data Set):**

\[ DS_r := \text{instr}(DS_2, "o") \]

would give error because DS_2 has more Measures.

---

**String length**

**Syntax**

\[
\text{length} ( \text{op} )
\]

**Input Parameters**

\[
\text{op} \quad \text{the operand}
\]

**Examples of valid syntaxes**

- length("Hello, World!")
- length(DS_1)

**Semantics for scalar operations**

Returns the length of a string. For example, length("Hello, World!") gives 13

For the empty string "" the value 0 is returned

**Input Parameters type**

\[
\text{op} ::
\begin{array}{c}
\text{dataset} \{ \text{measure<string>} \} \\
\mid \text{component<string>}
\end{array}
\]

**Result type**

\[
\text{result} ::
\begin{array}{c}
\text{dataset} \{ \text{measure<integer[value >= 0]> } \text{int_var} \} \\
\mid \text{component<integer[value >= 0]>}
\end{array}
\]

**Additional constraints**

For operations at Data Set level, the input Data Set must have exactly one string type Measure.

**Behaviour**
The operator has the behaviour of the "Operators changing the data type" (see the section "Typical behaviours of the ML Operators").

If \( op \) is a Data Set then **length** returns a dataset with a single measure \( \text{int\_var} \) of type integer.

**Examples**

Given the Data Set DS_1

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>null</td>
</tr>
</tbody>
</table>

**Example 1:** DS_r := length(DS_1) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>int_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>5</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>null</td>
</tr>
</tbody>
</table>

**Example 2 (on component):** DS_r := DS_1[calc Me_2:=length(Me_1)] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
<td>5</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>null</td>
<td>null</td>
</tr>
</tbody>
</table>

Given the Data Set DS_2:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;</td>
<td>&quot;world&quot;</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>null</td>
<td>&quot;hi&quot;</td>
</tr>
</tbody>
</table>

**Example 3 (applying the length operator at component level to a multi Measure Data Set):**

DS_r := DS_2 [calc Me_10:=length(Me_1), Me_20:=length(Me_2)] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_10</th>
<th>Me_20</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>&quot;hello&quot;&quot;world&quot;</td>
<td>5</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>null</td>
<td>&quot;hi&quot;</td>
<td>null</td>
<td>2</td>
</tr>
</tbody>
</table>

**Example 4 (length operator applied at Data Set level to a multi Measure Data Set):**

DS_r := length(DS_2) would give error because DS_2 has more Measures.
Unary plus: +

Syntax
+ op

Input parameters
op the operand

Examples of valid syntaxes
+ DS_1
+ 3

Semantics for scalar operations
The operator + returns the operand unchanged. For example:
+ 3 gives 3
+( - 5 ) gives - 5

Input Parameters type
op :: dataset { measure<number> _+ }
| component<number>
| number

Result type
result :: dataset { measure<number> _+ }
| component<number>
| number

Additional constraints
None.

Behaviour
The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).
According to the general rules about data types, the operator can be applied also on sub-types of number, that is the type integer. If the type of the operand is integer then the result has type integer. If the type of the operand is number then the result has type number.

Examples
Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>1.0</td>
<td>5</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>2.3</td>
<td>10</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>3.2</td>
<td>12</td>
</tr>
</tbody>
</table>

Example 1:
DS_r := + DS_1

results in:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>1.0</td>
<td>5</td>
</tr>
</tbody>
</table>
Example 2 (on components):
\[ DS_r := DS_1 [calc \; Me_3 := + Me_1 ] \] results in:

\[
\begin{array}{|c|c|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 & \text{Me}_2 & \text{Me}_3 \\
\hline
10 & A & 1.0 & 5 & 1.0 \\
10 & B & 2.3 & 10 & 2.3 \\
11 & A & 3.2 & 12 & 3.2 \\
\hline
\end{array}
\]

 Unary minus: \(-\)

Syntax
- op

Input parameters
op the operand

Examples of valid syntaxes
- DS_1
- 3

Semantics for scalar operations
The operator - inverts the sign of op. For example:
- 3 gives -3
- (-5) gives 5

Input Parameters type
op :: dataset \{ measure<
number> + \} | component<number> | number

Result type
result :: dataset \{ measure<number> + \} | component<number> | number

Additional constraints
None.

Behaviour
The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

According to the general rules about data types, the operator can be applied also on sub-types of number, that is the type integer. If the type of the operand is integer then the result has type integer. If the type of the operand is number then the result has type number.

Examples
Given the operand Data Set DS_1:

\[
\begin{array}{|c|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 & \text{Me}_2 \\
\hline
10 & A & 1 & 5.0 \\
\hline
\end{array}
\]
Example 1: \( DS_r := -DS_1 \) results in:

<table>
<thead>
<tr>
<th></th>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>-1</td>
<td>-5.0</td>
<td></td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>-2</td>
<td>-10.0</td>
<td></td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-3</td>
<td>-12.0</td>
<td></td>
</tr>
</tbody>
</table>

Example 2 (on components): \( DS_r := DS_1 \[ \text{calc Me}_3 := -Me_1 \] \) results in:

<table>
<thead>
<tr>
<th></th>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_3</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>1</td>
<td>5.0</td>
<td></td>
<td>-1</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>2</td>
<td>10.0</td>
<td></td>
<td>-2</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>3</td>
<td>12.0</td>
<td></td>
<td>-3</td>
</tr>
</tbody>
</table>

**Addition:**

**Syntax**

\( op1 + op2 \)

**Input parameters**

- \( op1 \) the first addend
- \( op2 \) the second addend

**Examples of valid syntaxes**

- \( DS_1 + DS_2 \)
- \( 3 + 5 \)

**Semantics for scalar operations**

The operator addition returns the sum of two numbers. For example:

\( 3 + 5 \) gives 8

**Input parameters type**

\( op1, op2 ::= \) dataset \{ measure<number> \_+ \}

| component<number> |
| number |

**Result type**

\( result ::= \) dataset \{ measure<number> \_+ \}

| component<number> |
| number |

**Additional constraints**

None.
**Behaviour**

The operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components” (see the section “Typical behaviours of the ML Operators”).

According to the general rules about data types, the operator can be applied also on sub-types of `number`, that is the type `integer`. If the type of both operands is `integer` then the result has type `integer`. If one of the operands is of type `number`, then the other operand is implicitly cast to `number` and therefore the result has type `number`.

**Examples**

Given the operand Data Sets `DS_1` and `DS_2`:

<p>| | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
<td>Id_2</td>
<td>Me_1</td>
<td>Me_2</td>
</tr>
<tr>
<td>10</td>
<td>A</td>
<td>5</td>
<td>5.0</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>2</td>
<td>10.5</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>3</td>
<td>12.2</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>4</td>
<td>20.3</td>
</tr>
</tbody>
</table>

**Example 1:**

```
DS_r := DS_1 + DS_2
```

results in:

<p>| | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
<td>Id_2</td>
<td>Me_1</td>
<td>Me_2</td>
</tr>
<tr>
<td>10</td>
<td>A</td>
<td>15</td>
<td>8.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>10</td>
<td>27.3</td>
</tr>
</tbody>
</table>

**Example 2:**

```
DS_r := DS_1 + 3
```

results in:

<p>| | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
<td>Id_2</td>
<td>Me_1</td>
<td>Me_2</td>
</tr>
<tr>
<td>10</td>
<td>A</td>
<td>8</td>
<td>8.0</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>5</td>
<td>13.5</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>6</td>
<td>15.2</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>7</td>
<td>23.3</td>
</tr>
</tbody>
</table>

**Example 3 (on components):**

```
DS_r := DS_1 [ calc Me_3 := Me_1 + 3.0 ]
```

results in:

<p>| | | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
<td>Id_2</td>
<td>Me_1</td>
<td>Me_2</td>
<td>Me_3</td>
</tr>
<tr>
<td>10</td>
<td>A</td>
<td>5</td>
<td>5.0</td>
<td>8.0</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>2</td>
<td>10.5</td>
<td>5.0</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>3</td>
<td>12.2</td>
<td>6.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>4</td>
<td>20.3</td>
<td>7.0</td>
</tr>
</tbody>
</table>
Subtraction:

**Syntax**

op1 - op2

**Input Parameters**

op1  the minuend
op2  the subtrahend

**Examples of valid syntaxes**

DS_1 - DS_2
3 - 5

**Semantics for scalar operations**

The operator subtraction returns the difference of two numbers. For example:

3 - 5  gives  -2

**Input Parameters type**

op1, op2::

| dataset { measure<number> _+ } |
| component<number> |
| number |

**Result type**

result::

| dataset { measure<number> _+ } |
| component<number> |
| number |

**Additional constraints**

None.

**Behaviour**

The operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components” (see the section “Typical behaviours of the ML Operators”).

According to the general rules about data types, the operator can be applied also on sub-types of number, that is the type integer. If the type of both operands is integer then the result has type integer. If one of the operands is of type number, then the other operand is implicitly cast to number and therefore the result has type number.

**Examples**

Given the operand Data Sets DS_1 and DS_2:

<table>
<thead>
<tr>
<th>DS_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td>-------</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>11</td>
</tr>
<tr>
<td>11</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>DS_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td>-------</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>11</td>
</tr>
</tbody>
</table>

**Example 1:**

DS_r := DS_1 - DS_2  results in:
Example 2: $DS_r := DS_1 - 3$ results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>-5</td>
<td>2.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>-2</td>
<td>13.3</td>
</tr>
</tbody>
</table>

Example 3 (on components): $DS_r := DS_1 [\text{calc} Me_3 := Me_1 - 3]$ results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_3</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>2</td>
<td>2.0</td>
<td>2</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>-1</td>
<td>7.5</td>
<td>-1</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>0</td>
<td>9.2</td>
<td>0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>1</td>
<td>17.3</td>
<td>1</td>
</tr>
</tbody>
</table>

Multiplication: $*$

**Syntax**

$\text{op1} * \text{op2}$

**Input parameters**

$\text{op1}$ the multiplicand

$\text{op2}$ the multiplier

**Examples of valid syntaxes**

$DS_1 * DS_2$

$3 * 5$

**Semantics for scalar operations**

The operator multiplication returns the product of two numbers. For example:

$3 * 5$ gives $15$

**Input parameters type**

$\text{op1, op2::dataset \{\text{measure<number> _+}\}}$

$| \text{component<number>}$

$| \text{number}$

**Result type**

$\text{result::dataset \{\text{measure<number> _+}\}}$

$| \text{component<number>}$

$| \text{number}$
Additional constraints

None.

 Behaviour

The operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components” (see the section “Typical behaviours of the ML Operators”).

According to the general rules about data types, the operator can be applied also on sub-types of number, that is the type integer. If the type of both operands is integer then the result has type integer. If one of the operands is of type number, then the other operand is implicitly cast to number and therefore the result has type number.

Examples

Given the operand Data Sets DS_1 and DS_2:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>100</td>
<td>7.6</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>10</td>
<td>12.3</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>20</td>
<td>25.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>2</td>
<td>20.0</td>
</tr>
</tbody>
</table>

DS_2

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>1</td>
<td>2.0</td>
</tr>
<tr>
<td>10</td>
<td>C</td>
<td>5</td>
<td>3.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>2</td>
<td>1.0</td>
</tr>
</tbody>
</table>

Example 1:  
DS_r := DS_1 * DS_2  
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>100</td>
<td>15.2</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>4</td>
<td>20.0</td>
</tr>
</tbody>
</table>

Example 2:  
DS_r := DS_1 * -3  
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>-300</td>
<td>-22.8</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>-30</td>
<td>-36.9</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-60</td>
<td>-75.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>-6</td>
<td>-60.0</td>
</tr>
</tbody>
</table>

Example 3 (on components):  
DS_r := DS_1 [ calc Me_3 := Me_1 * Me_2 ]  
results in:
Division:\n\n**Syntax**
\n\[ op1 \div op2 \]

**Input parameters**
- `op1`  the dividend
- `op2`  the divisor

**Examples of valid syntaxes**
- `DS_1 / DS_2`
- `3 / 5`

**Semantics for scalar operations**
- The operator `division` divides two numbers. For example:
  \[ \frac{3}{5} \] gives 0.6

**Input parameters type**
- `op1, op2`:
  - dataset { measure<number> _+ }
  - component<number>
  - number

**Result type**
- `result`:
  - dataset { measure<number> _+ }
  - component<number>
  - number

**Additional constraints**
- None.

**Behaviour**
- The operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components” (see the section “Typical behaviours of the ML Operators”).
- According to the general rules about data types, the operator can be applied also on sub-types of `number`, that is the type `integer`. The result has type `number`.
- If `op2` is 0 then the operation generates a run-time error.

**Examples**
- Given the operand Data Sets `DS_1` and `DS_2`:

```
<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>100</td>
<td>7.6</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>10</td>
<td>12.3</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>20</td>
<td>25.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>2</td>
<td>20.0</td>
</tr>
</tbody>
</table>
```
<table>
<thead>
<tr>
<th>Id</th>
<th>Id</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>1</td>
<td>2.0</td>
</tr>
<tr>
<td>10</td>
<td>C</td>
<td>2</td>
<td>1.0</td>
</tr>
</tbody>
</table>

**Example 1:**
\[
DS_r := DS_1 / DS_2
\]
results in:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>100</td>
<td>3.8</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>10</td>
<td>25.0</td>
</tr>
</tbody>
</table>

**Example 2:**
\[
DS_r := DS_1 / 10
\]
results in:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>10</td>
<td>0.76</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>1</td>
<td>1.23</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>2</td>
<td>2.5</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>0.2</td>
<td>2.0</td>
</tr>
</tbody>
</table>

**Example 3 (on components):**
\[
DS_r := DS_1 [ \text{calc} \ Me_3 := Me_2 / Me_1 ]
\]
results in:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_3</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>100</td>
<td>7.6</td>
<td>0.076</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>10</td>
<td>12.3</td>
<td>1.23</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>20</td>
<td>25.0</td>
<td>1.25</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>2</td>
<td>20.0</td>
<td>10.0</td>
</tr>
</tbody>
</table>

**Modulo:**

\[
\text{mod}
\]

**Syntax**

\[
\text{mod} \ (\text{op1}, \text{op2})
\]

**Input parameters**
- \text{op1} the dividend
- \text{op2} the divisor

**Examples of valid syntaxes**
The operator mod returns the remainder of \( op1 \) divided by \( op2 \). It returns \( op1 \) if divisor \( op2 \) is 0. For example:

\[
\begin{align*}
\text{mod} \ (5, 2) & \quad \text{gives 1} \\
\text{mod} \ (5, -2) & \quad \text{gives -1} \\
\text{mod} \ (8, 2) & \quad \text{gives 0} \\
\text{mod} \ (9, 0) & \quad \text{gives 9}
\end{align*}
\]

**Input Parameters type**

\( op1, op2 :: \) dataset \{ measure\langle number\rangle \_+ \}

\| component\langle number\rangle
\| number

**Result type**

\( result :: \) dataset \{ measure\langle number\rangle \_+ \}

\| component\langle number\rangle
\| number

**Additional constraints**

None.

**Behaviour**

The operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components” (see the section “Typical behaviours of the ML Operators”).

According to the general rules about data types, the operator can be applied also on sub-types of \textit{number}, that is the type \textit{integer}. If the type of both operands is \textit{integer} then the result has type \textit{integer}. If one of the operands is of type \textit{number}, then the other operand is implicitly cast to \textit{number} and therefore the result has type \textit{number}.

**Examples**

Given the operand Data Sets DS\(_1\) and DS\(_2\):

\[
\begin{array}{|c|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 & \text{Me}_2 \\
\hline
10 & A & 100 & 0.7545 \\
10 & B & 10 & 18.45 \\
11 & A & 20 & 1.87 \\
11 & B & 9 & 12.3 \\
\hline
\end{array}
\]

\[
\begin{array}{|c|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 & \text{Me}_2 \\
\hline
10 & A & 1 & 0.25 \\
10 & C & 5 & 3.0 \\
11 & B & 2 & 2.0 \\
\hline
\end{array}
\]

**Example 1:**

\[
\text{DS}_r := \text{mod} \ (\text{DS}_1, \text{DS}_2)
\]

results in:
Example 2: \[ \text{DS}_r := \text{mod}(\text{DS}_1, 15) \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>10</td>
<td>0.7545</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>10</td>
<td>3.45</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>5</td>
<td>1.87</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>9</td>
<td>12.3</td>
</tr>
</tbody>
</table>

Example 3 (on components): \[ \text{DS}_r := \text{DS}_1[ \text{calc} \ Me_3 := \text{mod}(\text{DS}_1\#\text{Me}_1, 3.0)] \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_3</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>100</td>
<td>0.7545</td>
<td>1.0</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>10</td>
<td>18.45</td>
<td>1.0</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>20</td>
<td>1.87</td>
<td>2.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>9</td>
<td>12.3</td>
<td>0.0</td>
</tr>
</tbody>
</table>

Rounding: \textbf{round}

\textit{Syntax}

\texttt{round ( op, numDigit )}

\textit{Input parameters}

\texttt{op} the operand

\texttt{numDigit} the number of positions to round to

\textit{Examples of valid syntaxes}

\texttt{round ( DS_1 , 2 )}

\texttt{round ( DS_2 )}

\texttt{round ( 3.14159 , 2 )}

\texttt{round ( 3.14159 , _ )}

\textit{Semantics for scalar operations}

The operator \texttt{round} rounds the operand to a number of positions at the right of the decimal point equal to the \texttt{numDigit} parameter. The decimal point is assumed to be at position 0. If \texttt{numDigit} is negative, the rounding happens at the left of the decimal point. The rounding operation leaves the \texttt{numDigit} position unchanged if the \texttt{numDigit}+1 position is between 0 and 4, otherwise it adds 1 to the number that is in the \texttt{numDigit} position. All the positions greater than \texttt{numDigit} are set to 0. The basic scalar type of the result is \texttt{integer} if \texttt{numDigit} is omitted, \texttt{number} otherwise.

For example:

\texttt{round ( 3.14159 , 2 )} gives 3.14

\texttt{round ( 3.14159 , 4 )} gives 3.1416

\texttt{round ( 12345.6 , 0 )} gives 12346.0
round ( 12345.6 ) gives 12346
round ( 12345.6, _ ) gives 12346
round ( 12345.6, -1 ) gives 12350.0

Input parameters type
op1 :: dataset { measure<number> _+ } |
| component<number>
| number
numDigit:: component < integer > |
| integer

Result type
result :: dataset { measure<number> _+ } |
| component<number>
| number

Additional constraints
None.

Behaviour
As for the invocations at Data Set level, the operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component”, as for the invocations at Component or Scalar level, the operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components”, (see the section “Typical behaviours of the ML Operators”).

Examples
Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_1</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.5</td>
<td>5.9</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>36.2</td>
<td>17.7</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.5</td>
<td>24.3</td>
</tr>
</tbody>
</table>

Example 1: DS_r := round(DS_1, 0) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>8.0</td>
<td>6.0</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>7.0</td>
<td>6.0</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>36.0</td>
<td>18.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>45.0</td>
<td>24.0</td>
</tr>
</tbody>
</table>

Example 2 (on components): DS_r := DS_1 [ calc Me_10 := round ( Me_1 ) ] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_10</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.5</td>
<td>5.9</td>
<td>8</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>7.1</td>
<td>5.5</td>
<td>7</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>36.2</td>
<td>17.7</td>
<td>36</td>
</tr>
</tbody>
</table>
Example 3 (on components): \[ DS_r := DS_1 \ [ \text{calc}\ Me_20 := \text{round}( Me_1 , -1 ) ] \] results in:

<table>
<thead>
<tr>
<th>DS_r</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td>------</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>11</td>
</tr>
<tr>
<td>11</td>
</tr>
</tbody>
</table>

Truncation: \( \text{trunc} \)

**Syntax**

\[ \text{trunc}(\ op,\ numDigit) \]

**Input Parameters**

- \( \text{op} \): the operand
- \( \text{numDigit} \): the number of position from which to trunc

**Examples of valid syntaxes**

- \( \text{trunc}( DS_1, 2 ) \)
- \( \text{trunc}( DS_1 ) \)
- \( \text{trunc}( 3.14159, 2 ) \)
- \( \text{trunc}( 3.14159, _ ) \)
- \( \text{trunc}( 12345.6, _ ) \)

**Semantics for scalar operations**

The operator \( \text{trunc} \) truncates the operand to a number of positions at the right of the decimal point equal to the \( \text{numDigit} \) parameter. The decimal point is assumed to be at position 0. If \( \text{numDigit} \) is negative, the truncation happens at the left of the decimal point. The truncation operation leaves the \( \text{numDigit} \) position unchanged. All the positions greater than \( \text{numDigit} \) are eliminated. The basic scalar type of the result is \( \text{integer} \) if \( \text{numDigit} \) is omitted, \( \text{number} \) otherwise.

For example:

- \( \text{trunc}( 3.14159, 2 ) \) gives 3.14
- \( \text{trunc}( 3.14159, 4 ) \) gives 3.1415
- \( \text{trunc}( 12345.6, 0 ) \) gives 12345.0
- \( \text{trunc}( 12345.6 ) \) gives 12345
- \( \text{trunc}( 12345.6, _ ) \) gives 12345
- \( \text{trunc}( 12345.6, -1 ) \) gives 12340.0

**Input parameters type**

- \( \text{op} ::= \) dataset \{ measure<number> \_+ \} | component<number> | number
- \( \text{numDigit} ::= \) component < integer > | integer

**Result type**

- \( \text{result} ::= \) dataset \{ measure<number> \_+ \} | component<number> | number

**Additional constraints**

None.
**Behaviour**

As for the invocations at Data Set level, the operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component”, as for the invocations at Component or Scalar level, the operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components”, (see the section “Typical behaviours of the ML Operators”).

**Examples**

Given the operand Data Set DS_1:

<p>| | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.5</td>
<td>5.9</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>7.1</td>
<td>5.5</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>36.2</td>
<td>17.7</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.5</td>
<td>24.3</td>
</tr>
</tbody>
</table>

**Example 1:**

\[ DS_r := \text{trunc}(DS_1, 0) \]

results in:

<p>| | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.0</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>7.0</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>36.0</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.0</td>
</tr>
</tbody>
</table>

**Example 2 (on components):**

\[ DS_r := DS_1[ \text{calc } Me_10 := \text{trunc}(Me_1) ] \]

results in:

<p>| | | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.5</td>
<td>5.9</td>
<td>7</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>7.1</td>
<td>5.5</td>
<td>7</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>36.2</td>
<td>17.7</td>
<td>36</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.5</td>
<td>24.3</td>
<td>44</td>
</tr>
</tbody>
</table>

**Example 3 (on components):**

\[ DS_r := DS_1[ \text{calc } Me_20 := \text{trunc}(Me_1, -1) ] \]

results in:

<p>| | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.5</td>
<td>5.9</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>7.1</td>
<td>5.5</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>36.2</td>
<td>17.7</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.5</td>
<td>24.3</td>
</tr>
</tbody>
</table>
Ceiling: \( \text{ceil} \)

**Syntax**

\[
\text{ceil} \ ( \text{op} )
\]

**Input parameters**

- op the operand

**Examples of valid syntaxes**

- \( \text{ceil} \ ( \text{DS}_1 \) \)
- \( \text{ceil} \ ( 3.14159 \) \)

**Semantics for scalar operations**

The operator \( \text{ceil} \) returns the smallest integer greater than or equal to \( \text{op} \).

For example:

- \( \text{ceil}(3.14159) \) gives \( 4 \)
- \( \text{ceil}(15) \) gives \( 15 \)
- \( \text{ceil}(-3.1415) \) gives \( -3 \)
- \( \text{ceil}(-0.1415) \) gives \( 0 \)

**Input parameters type**

- \( \text{op} :: \) dataset \{ measure<number> \} +
  | component<number>
  | number

**Result type**

- \( \text{result} :: \) dataset \{ measure<integer> \} +
  | component<integer>
  | integer

**Additional constraints**

None.

**Behaviour**

The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

**Examples**

Given the operand Data Set \( \text{DS}_1 \):

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_1</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.0</td>
<td>5.9</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>0.1</td>
<td>-5.0</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-32.2</td>
<td>17.7</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.5</td>
<td>-0.3</td>
</tr>
</tbody>
</table>

Example 1: \( \text{DS}_r := \text{ceil} \ ( \text{DS}_1 \) \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_1</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7</td>
<td>6</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>1</td>
<td>-5</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-32</td>
<td>18</td>
</tr>
</tbody>
</table>
Example 2 (on components):

\[
\begin{align*}
\text{DS}_r & := \text{DS}_1 \{ \text{calc Me}_{10} := \text{ceil Me}_1 \}
\end{align*}
\]
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_1</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_{10}</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.0</td>
<td>5.9</td>
<td>7</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>0.1</td>
<td>-5.0</td>
<td>1</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-32.2</td>
<td>17.7</td>
<td>-32</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.5</td>
<td>-0.3</td>
<td>45</td>
</tr>
</tbody>
</table>

**Floor:**

**Syntax**

\[
\text{floor} \ ( \text{op} )
\]

**Input parameters**

<table>
<thead>
<tr>
<th>op</th>
<th>the operand</th>
</tr>
</thead>
</table>

**Examples of valid syntaxes**

- floor (DS_1)
- floor (3.14159)

**Semantics for scalar operations**

The operator **floor** returns the greatest integer which is smaller than or equal to **op**.

For example:

- floor (3.1415) gives 3
- floor (15) gives 15
- floor (-3.1415) gives -4
- floor (-0.1415) gives -1

**Input parameters type**

<table>
<thead>
<tr>
<th>op ::</th>
<th>dataset { measure&lt;number&gt; _+ }</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>component&lt;number&gt;</td>
</tr>
<tr>
<td></td>
<td>number</td>
</tr>
</tbody>
</table>

**Result type**

<table>
<thead>
<tr>
<th>result ::</th>
<th>dataset { measure&lt;integer&gt; _+ }</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>component&lt;integer&gt;</td>
</tr>
<tr>
<td></td>
<td>integer</td>
</tr>
</tbody>
</table>

**Additional constraints**

None.

**Behaviour**

The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

**Examples**

Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_1</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>11</td>
<td>B</td>
<td>45</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td></td>
<td>7.0</td>
<td>5.9</td>
</tr>
<tr>
<td>----</td>
<td>----</td>
<td>-----</td>
<td>-----</td>
</tr>
<tr>
<td>10</td>
<td>A</td>
<td>7.0</td>
<td>5.9</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>0.1</td>
<td>-5.0</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-32.2</td>
<td>17.7</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.5</td>
<td>-0.3</td>
</tr>
</tbody>
</table>

**Example 1:**  
\[ DS_r := \text{floor}( DS_1 ) \]
results in:

<p>| | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>0</td>
<td>-5</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-33</td>
<td>17</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44</td>
<td>-1</td>
</tr>
</tbody>
</table>

**Example 2 (on components):**  
\[ DS_r := DS_1[ \text{calc} \ Me_10 := \text{floor}(Me_1) ] \]
results in:

<p>| | | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>7.5</td>
<td>5.9</td>
<td>7</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>0.1</td>
<td>-5.5</td>
<td>0</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-32.2</td>
<td>17.7</td>
<td>-33</td>
</tr>
<tr>
<td>11</td>
<td>B</td>
<td>44.5</td>
<td>-0.3</td>
<td>44</td>
</tr>
</tbody>
</table>

**Absolute value : \( \textbf{abs} \)**

**Syntax**
\[
\textbf{abs} \ ( \text{op} )
\]

**Input parameters**
- \( \text{op} \): the operand

**Examples of valid syntaxes**
- \( \text{abs}( DS_1 ) \)
- \( \text{abs}(-5) \)

**Semantics for scalar operations**
The operator \( \textbf{abs} \) calculates the absolute value of a number.
For example:
- \( \text{abs}(-5.49) \) gives 5.49
- \( \text{abs}(5.49) \) gives 5.49

**Input parameters type**
- \( \text{op} :: \) dataset { measure<number> + }
  - component<number>
  - number

**Result type**
- \( \text{result} :: \) dataset { measure<number [ value >= 0 ]> + }
  - component<number [ value >= 0 ]>
| number [ value >= 0 ]

Additional constraints
None.

Behaviour
The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

Examples
Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>0.484183</td>
<td>0.7545</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>-0.515817</td>
<td>-13.45</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-1.000000</td>
<td>187.0</td>
</tr>
</tbody>
</table>

Example 1: \( DS_r := \text{abs} (\ DS_1) \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>0.484183</td>
<td>0.7545</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>0.515817</td>
<td>13.45</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>1.000000</td>
<td>187</td>
</tr>
</tbody>
</table>

Example 2 (on components): \( DS_r := \ DS_1 [\ calc\ Me_10 := \text{abs}(\ Me_1) ] \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
<th>Me_10</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>0.484183</td>
<td>0.7545</td>
<td>0.484183</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>-0.515817</td>
<td>-13.45</td>
<td>0.515817</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>-1.000000</td>
<td>187</td>
<td>1.000000</td>
</tr>
</tbody>
</table>

Exponential: \( \text{exp} \)

Syntax
\( \text{exp} (\ op) \)

Input parameters
op the operand

Examples of valid syntaxes
\( \text{exp} (\ DS_1) \)
\( \text{exp} (\ 5) \)

Semantics for scalar operations
The operator \( \text{exp} \) returns \( e \) (base of the natural logarithm) raised to the op-th power.

For example:
\( \text{exp} (\ 5) \) gives 148.41315...
\( \text{exp} (\ 1) \) gives 2.71828... (the number \( e \))
\( \exp(0) \) gives 1.0

\( \exp(-1) \) gives 0.36787… (the number 1/e)

**Input parameters type**

\( \text{op} :: \) dataset \( \{ \text{measure<number> _+} \}

\| component<number>

\| number

**Result type**

\( \text{result} :: \) dataset \( \{ \text{measure<number [value > 0]> _+} \}

\| component<number [value > 0]>

\| number[value > 0]

**Additional constraints**

None.

**Behaviour**

The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

**Examples**

Given the operand Data Set \( DS_1 \):

<table>
<thead>
<tr>
<th>( DS_1 )</th>
<th>( Id_1 )</th>
<th>( Id_2 )</th>
<th>( Me_1 )</th>
<th>( Me_2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>10</td>
<td>A</td>
<td>5</td>
<td>0.7545</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>B</td>
<td>8</td>
<td>13.45</td>
</tr>
<tr>
<td></td>
<td>11</td>
<td>A</td>
<td>2</td>
<td>1.87</td>
</tr>
</tbody>
</table>

**Example 1:** \( DS_r := \exp(DS_1) \) results in:

<table>
<thead>
<tr>
<th>( DS_r )</th>
<th>( Id_1 )</th>
<th>( Id_2 )</th>
<th>( Me_1 )</th>
<th>( Me_2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>10</td>
<td>A</td>
<td>148.413</td>
<td>2.126547</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>B</td>
<td>2980.95</td>
<td>693842.3</td>
</tr>
<tr>
<td></td>
<td>11</td>
<td>A</td>
<td>7.38905</td>
<td>6.488296</td>
</tr>
</tbody>
</table>

**Example 2 (on components):** \( DS_r := DS_1 \ [ \text{calc Me}_1 := \exp(\text{Me}_1) ] \) results in:

<table>
<thead>
<tr>
<th>( DS_r )</th>
<th>( Id_1 )</th>
<th>( Id_2 )</th>
<th>( Me_1 )</th>
<th>( Me_2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>10</td>
<td>A</td>
<td>148.413</td>
<td>0.7545</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>B</td>
<td>2980.95</td>
<td>13.45</td>
</tr>
<tr>
<td></td>
<td>11</td>
<td>A</td>
<td>7.389</td>
<td>1.87</td>
</tr>
</tbody>
</table>

**Natural logarithm:** \( \ln \)

**Syntax**

\( \ln ( \text{op} ) \)
Input parameters

\textbf{op} the operand

Examples of valid syntaxes

\texttt{ln ( DS_1 )}
\texttt{ln ( 148 )}

Semantics for scalar operations

The operator \texttt{ln} calculates the natural logarithm of a number.

For example:

\begin{itemize}
  \item \texttt{ln ( 148 )} gives 4.997…
  \item \texttt{ln ( e )} gives 1.0
  \item \texttt{ln ( 1 )} gives 0.0
  \item \texttt{ln ( 0.5 )} gives -0.693…
\end{itemize}

Input parameters type

\begin{verbatim}
op :: dataset { measure<number [value > 0] > } |
    component<number [value > 0] > |
    number
\end{verbatim}

Result type

\begin{verbatim}
result :: dataset { measure<number > } |
    component<number > |
    number
\end{verbatim}

Additional constraints

None.

Behaviour

The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

Examples

Given the operand Data Set DS_1:

\begin{verbatim}
DS_1
\end{verbatim}

\begin{tabular}{|c|c|c|c|}
\hline
Id_1 & Id_2 & Me_1 & Me_2 \\
\hline
10   & A    & 148.413 & 0.7545 \\
10   & B    & 2980.95 & 13.45  \\
11   & A    & 7.38905 & 1.87   \\
\hline
\end{tabular}

Example 1:

\begin{verbatim}
DS_r := ln(DS_1)
\end{verbatim}

results in:

\begin{verbatim}
DS_r
\end{verbatim}

\begin{tabular}{|c|c|c|c|}
\hline
Id_1 & Id_2 & Me_1 & Me_2 \\
\hline
10   & A    & 5.0  & -0.281700 \\
10   & B    & 8.0  & 2.598979  \\
11   & A    & 2.0  & 0.625938  \\
\hline
\end{tabular}

Example 2 (on components):

\begin{verbatim}
DS_r := DS_1 [ calc Me_2 := ln ( DS_1#Me_1 ) ]
\end{verbatim}

results in:
Power : \texttt{power}

\textit{Syntax}
\begin{verbatim}
  power ( base , exponent )
\end{verbatim}

\textit{Input parameters}
- \texttt{base} : the operand
- \texttt{exponent} : the exponent of the power

\textit{Examples of valid syntaxes}
- \texttt{power ( DS_1, 2 )}
- \texttt{power ( 5, 2 )}

\textit{Semantics for scalar operations}
The operator \texttt{power} raises a number (the \texttt{base}) to another one (the \texttt{exponent}). For example:
- \texttt{power ( 5, 2 )} gives 25
- \texttt{power ( 5, 1 )} gives 5
- \texttt{power ( 5, 0 )} gives 1
- \texttt{power ( 5, -1 )} gives 0.2
- \texttt{power ( -5, 3 )} gives -125

\textit{Input parameters type}
- \texttt{base} :: \texttt{dataset} \{ \texttt{measure<number>}_+ \}
- | \texttt{component<number>}
- | \texttt{number}

- \texttt{exponent} :: \texttt{component<number>}
- | \texttt{number}

\textit{Result type}
- \texttt{result} :: \texttt{dataset} \{ \texttt{measure<number>}_+ \}
- | \texttt{component<number>}
- | \texttt{number}

\textit{Additional constraints}
- None.

\textit{Behaviour}
As for the invocations at Data Set level, the operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component”, as for the invocations at Component or Scalar level, the operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components”, (see the section “Typical behaviours of the ML Operators”).

\textit{Examples}
Given the operand Data Set \texttt{DS_1}:

\begin{table}[h]
\centering
\begin{tabular}{|c|c|c|c|}
\hline
\texttt{Id_1} & \texttt{Id_2} & \texttt{Me_1} & \texttt{Me_2} \\
\hline
10 & A & 148.413 & 5.0 \\
10 & B & 2980.95 & 8.0 \\
11 & A & 7.38905 & 2.0 \\
\hline
\end{tabular}
\caption{Data Set DS_1}
\end{table}
Example 1: \[ DS_r := \text{power}(DS_1, 2) \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>9</td>
<td>0.56927</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>16</td>
<td>180.9025</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>25</td>
<td>3.4969</td>
</tr>
</tbody>
</table>

Example 2 (on components): \[ DS_r := DS_1[ \text{calc Me}_1 := \text{power(Me}_1, 2) ] \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>9</td>
<td>0.7545</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>16</td>
<td>13.45</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>25</td>
<td>1.87</td>
</tr>
</tbody>
</table>

Logarithm: \( \text{log} \)

Syntax

\[ \text{log} \ ( \text{op} , \text{num} ) \]

Input parameters

- \text{op} \quad \text{the base of the logarithm}
- \text{num} \quad \text{the number to which the logarithm is applied}

Examples of valid syntaxes

\[ \text{log} \ ( DS_1, 2 ) \]
\[ \text{log} \ ( 1024, 2 ) \]

Semantics for scalar operations

The operator \( \text{log} \) calculates the logarithm of \text{num} base \text{op}.

For example:

\[ \text{log} \ ( 1024, 2 ) \quad \text{gives 10} \]
\[ \text{log} \ ( 1024, 10 ) \quad \text{gives 3.01} \]

Input parameters type

- \text{op} :: dataset \{ \text{measure}<\text{number}[\text{value}>1]\}_+\}
- \text{num} :: component<\text{number}[\text{value}>0]>

Result type

- \text{result} :: dataset \{ \text{measure}<\text{number}>\}_+\}
- \text{number}
Additional constraints
None.

Behaviour
As for the invocations at Data Set level, the operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component”, as for the invocations at Component or Scalar level, the operator has the behaviour of the “Operators applicable on two Scalar Values or Data Sets or Data Set Components”, (see the section “Typical behaviours of the ML Operators”).

Examples
Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>1024</td>
<td>0.7545</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>64</td>
<td>13.45</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>32</td>
<td>1.87</td>
</tr>
</tbody>
</table>

Example 1:           DS_r := log ( DS_1, 2 ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>10.0</td>
<td>-0.40641</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>6.0</td>
<td>3.749534</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>5.0</td>
<td>0.903038</td>
</tr>
</tbody>
</table>

Example 2 (on components): DS_r := DS_1 [ calc Me_1 := log (Me_1, 2) ] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>10.0</td>
<td>0.7545</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>6.0</td>
<td>13.45</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>5.0</td>
<td>1.87</td>
</tr>
</tbody>
</table>

Square root:

Syntax
sqrt ( op )

Input parameters
op     the operand

Examples of valid syntaxes
sqrt ( DS_1 )
sqrt ( 5 )

Semantics for scalar operations
The operator sqrt calculates the square root of a number. For example:
sqrt ( 25 ) gives 5
**Input parameters type**

\[
op:: \text{dataset} \left\{ \text{measure<number [value >= 0] > + } \right\} \\
\quad | \text{component<number [value >= 0] >} \\
\quad | \text{number [value >= 0]} \]

**Result type**

\[
\text{result::} \text{dataset} \left\{ \text{measure<number [value >= 0] > + } \right\} \\
\quad | \text{component<number [value >= 0] >} \\
\quad | \text{number [value >= 0]} \]

**Additional constraints**

None.

**Behaviour**

The operator has the behaviour of the “Operators applicable on one Scalar Value or Data Set or Data Set Component” (see the section “Typical behaviours of the ML Operators”).

**Examples**

Given the operand Data Set \(DS_1\):

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>16</td>
<td>0.7545</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>81</td>
<td>13.45</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>64</td>
<td>1.87</td>
</tr>
</tbody>
</table>

**Example 1:**

\[DS_r := \sqrt{DS_1}\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>4</td>
<td>0.86862</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>9</td>
<td>3.667424</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>8</td>
<td>1.367479</td>
</tr>
</tbody>
</table>

**Example 2 (on components):**

\[DS_r := DS_1 [ \text{calc Me}_1 := \sqrt{Me_1} ]\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>A</td>
<td>4</td>
<td>0.7545</td>
</tr>
<tr>
<td>10</td>
<td>B</td>
<td>9</td>
<td>13.45</td>
</tr>
<tr>
<td>11</td>
<td>A</td>
<td>8</td>
<td>1.87</td>
</tr>
</tbody>
</table>
Equal to : 

\[ = \]

### Syntax
\[ \text{left} = \text{right} \]

### Input parameters
- **left**: the left operand
- **right**: the right operand

### Examples of valid syntaxes
- \[ \text{DS}_1 = \text{DS}_2 \]

### Semantics for scalar operations
- The operator returns **TRUE** if the left is equal to right, **FALSE** otherwise.
- For example:
  - \[ 5 = 9 \] gives: **FALSE**
  - \[ 5 = 5 \] gives: **TRUE**
  - \[ "hello" = "hi" \] gives: **FALSE**

### Input parameters type
- **left**, **right** :: dataset { measure<scalar> _ }  
  | component<scalar>  
  | scalar

### Result type
- **result** :: dataset { measure<boolean> bool_var }  
  | component<boolean>  
  | boolean

### Additional constraints
- Operands **left** and **right** must be of the same scalar type

### Behaviour
- The operator has the typical behaviour of the “Operators changing the data type” (see the section “Typical behaviours of the ML Operators”).

### Examples
- Given the operand Data Set **DS_1**:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>0.286</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>0.064</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>0.043</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>0.08</td>
</tr>
<tr>
<td>2012</td>
<td>W</td>
<td>Total</td>
<td>Total</td>
<td>0.08</td>
</tr>
</tbody>
</table>
Example 1: \[ DS_r := DS_1 = 0.08 \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>W</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Example 2 (on Components): \[ DS_r := DS_1 \ [ \text{calc} \ Me_2 := Me_1 = 0.08 \] \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>0.286</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>0.064</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>0.043</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>0.08</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>W</td>
<td>Total</td>
<td>Total</td>
<td>0.08</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Not equal to: \(<>\)

**Syntax**

\[ \text{left} <> \text{right} \]

**Input parameters**

- **left** the left operand
- **right** the right operand

**Examples of valid syntaxes**

- DS_1 <> DS_2
- DS_1 <> DS_2

**Semantics for scalar operations**

The operator returns FALSE if the left is equal to right, TRUE otherwise. For example:

- \(5 <> 9\) gives: TRUE
- \(5 <> 5\) gives: FALSE
- “hello” <> “hi” gives: TRUE

**Input parameters type**

- **left**, right :: dataset \{measure<scalar> \_\}
- | component<scalar>
- | scalar
Result type

result:: dataset { measure<boolean> bool_var }
| component<boolean>
| boolean

Additional constraints

Operands left and right must be of the same scalar type

Behaviour

The operator has the typical behaviour of the “Operators changing the data type” (see the section “Typical behaviours of the ML Operators”).

Examples

Given the operand Data Sets DS_1 and DS_2:

<table>
<thead>
<tr>
<th>DS_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td>G</td>
</tr>
<tr>
<td>R</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>DS_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td>G</td>
</tr>
<tr>
<td>R</td>
</tr>
</tbody>
</table>

Example 1: DS_r := DS_1 <> DS_2 results in:

<table>
<thead>
<tr>
<th>DS_r</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td>G</td>
</tr>
<tr>
<td>R</td>
</tr>
</tbody>
</table>

Note that due to the behaviour for NULL values, if the value for Greece in the second operand had also been NULL, then the result would still be NULL for Greece.

Example 2 (on Components): DS_r := DS_1 [ calc Me_2 := Me_1<>7.5 ] results in:

<table>
<thead>
<tr>
<th>DS_r</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td>G</td>
</tr>
<tr>
<td>R</td>
</tr>
</tbody>
</table>

Greater than : > >=

Syntax

left { > | >= } right
Input parameters
left  the left operand part of the comparison
right the right operand part of the comparison

Examples of valid syntaxes
DS_1 > DS_2
DS_1 >= DS_2

Semantics for scalar operations
The operator > returns TRUE if left is greater than right, FALSE otherwise.
The operator >= returns TRUE if left is greater than or equal to right, FALSE otherwise.

For example:
5 > 9 gives: FALSE
5 >= 5 gives: TRUE
“hello” > “hi” gives: FALSE

Input parameters type
left, right :: dataset { measure<scalar> _ }
| component<scalar>
| scalar

Result type
result :: dataset { measure<boolean> bool_var }
| component<boolean>
| boolean

Additional constraints
Operands left and right must be of the same scalar type

Behaviour
The operator has the typical behaviour of the “Operators changing the data type” (see the section “Typical behaviours of the ML Operators”).

Examples
Given the operand Data Set DS_1:

Example 1:  DS_r := DS_1 > 20 results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Id_5</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>G</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>NULL</td>
</tr>
<tr>
<td>2</td>
<td>R</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>12.2</td>
</tr>
<tr>
<td>2</td>
<td>F</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>29.5</td>
</tr>
</tbody>
</table>

Example 2 (on Components):  DS_r := DS_1 [ calc Me_2 := Me_1 > 20 ] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Id_5</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>G</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>NULL</td>
</tr>
<tr>
<td>2</td>
<td>R</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>FALSE</td>
</tr>
<tr>
<td>2</td>
<td>F</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>TRUE</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>---</td>
<td>---</td>
<td>---</td>
<td>---</td>
<td>---</td>
<td>---</td>
</tr>
<tr>
<td>2</td>
<td>G</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>NULL</td>
</tr>
<tr>
<td>2</td>
<td>R</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>12.2</td>
</tr>
<tr>
<td>2</td>
<td>F</td>
<td>2011</td>
<td>Total</td>
<td>Percentage</td>
<td>29.5</td>
</tr>
</tbody>
</table>

Given the left operand Data Set:

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>G</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td></td>
<td>7.1</td>
</tr>
<tr>
<td>R</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td></td>
<td>42.5</td>
</tr>
</tbody>
</table>

and the right operand Data Set:

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>G</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td></td>
<td>7.5</td>
</tr>
<tr>
<td>R</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td></td>
<td>33.7</td>
</tr>
</tbody>
</table>

Example 3: \( DS_r := DS_1 > DS_2 \)

If the Me_1 column for Germany in the DS_2 Data Set had a NULL value the result would be:

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>G</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td></td>
<td>FALSE</td>
</tr>
<tr>
<td>R</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td></td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Less than: \( < \) \( \leq \)

**Syntax**

\[
\text{left} \{ < | \leq \} \text{right}
\]

**Input parameters**

- left: the left operand
- right: the right operand

**Examples of valid syntaxes**

- \( DS_1 < DS_2 \)
- \( DS_1 \leq DS_2 \)

**Semantics for scalar operations**

- The operator \( < \) returns TRUE if left is smaller than right, FALSE otherwise.
- The operator \( \leq \) returns TRUE if left is smaller than or equal to right, FALSE otherwise.
For example:

- $5 < 4$ gives: FALSE
- $5 \leq 5$ gives: TRUE
- "hello" < "hi" gives: TRUE

**Input parameters type**

- $left, right :: \text{dataset}\ \{\text{measure<scalar> \_}\}$
- | component<scalar>
- | scalar

**Result type**

- $result :: \text{dataset}\ \{\text{measure<boolean> \_ boolean}\}$
- | component<boolean>
- | boolean

**Additional constraints**

Operands $left$ and $right$ must be of the same scalar type

** Behaviour**

The operator has the typical behaviour of the "Operators changing the data type" (see the section "Typical behaviours of the ML Operators").

**Examples**

Given the operand Data Set $DS_1$:

<table>
<thead>
<tr>
<th>$DS_1$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$id_1$</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
</tbody>
</table>

**Example 1**

$DS_r := DS_1 < 15000000$

results in:

<table>
<thead>
<tr>
<th>$DS_r$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$id_1$</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
</tbody>
</table>

**Between**

**Syntax**

$between$ (op, from, to)
Input parameters

\text{op} \quad \text{the Data Set to be checked}

\text{from} \quad \text{the left delimiter}

\text{to} \quad \text{the right delimiter}

Examples of valid syntaxes

\text{ds2 := between(ds1, 5,10)}

\text{ds2 := ds1 \ [\ calc \ m1 := between(me2, 5, 10) \ ]}

Semantics for scalar operations

The operator returns \text{TRUE} if \text{op} is greater than or equal to \text{from} and lower than or equal to \text{to}. In other terms, it is a shortcut for the following:

\text{op} \geq \text{from} \text{ and } \text{op} \leq \text{to}

The types of \text{op}, \text{from} and \text{to} must be compatible scalar types.

Input parameters type

\begin{align*}
\text{op} & : \quad \text{dataset \ \{} \text{measure<scalar>} \_ \\
& | \quad \text{component<scalar>} \\
& | \quad \text{scalar}
\end{align*}

\begin{align*}
\text{from} & : \quad \text{scalar} \ | \ \text{component<scalar>}
\end{align*}

\begin{align*}
\text{to} & : \quad \text{scalar} \ | \ \text{component<scalar>}
\end{align*}

Result type

\begin{align*}
\text{result} & : \quad \text{dataset \ \{} \ \text{measure<boolean>} \ \text{bool_var} \ \}
\end{align*}

\begin{align*}
& | \quad \text{component<boolean>} \\
& | \quad \text{boolean}
\end{align*}

Additional constraints

The type of the operand (i.e., the measure of the dataset, the type of the component, the scalar type) must be the same as that of from and to.

Behaviour

The operator has the typical behaviour of the “Operators changing the data type” (see the section “Typical behaviours of the ML Operators”).

Examples

Given the following Data Set \text{DS_1}:

\begin{tabular}{|c|c|c|c|c|}
\hline
\text{DS_1} & \text{Id_1} & \text{Id_2} & \text{Id_3} & \text{Id_4} & \text{Me_1} \\
\hline
G & Total & Percentage & Total & 6 \\
R & Total & Percentage & Total & -2 \\
\hline
\end{tabular}

Example 1: \quad \text{DS_r:= between(ds1, 5,10)} \quad \text{results in:}

\begin{tabular}{|c|c|c|c|c|c|}
\hline
\text{DS_1} & \text{Id_1} & \text{Id_2} & \text{Id_3} & \text{Id_4} & \text{bool_var} \\
\hline
G & Total & Percentage & Total & TRUE \\
R & Total & Percentage & Total & FALSE \\
\hline
\end{tabular}
Element of: \( \text{in} / \text{not_in} \)

**Syntax**

\[
\begin{align*}
\text{op in \ collection} \\
\text{op not_in \ collection}
\end{align*}
\]

\[
\text{collection ::= set} \mid \text{valueDomainName}
\]

**Input parameters**

- **op**: the operand to be tested
- **collection**: the Set or the Value Domain which contains the values
- **set**: the Set which contains the values (it can be a Set name or a Set literal)
- **valueDomainName**: the name of the Value Domain which contains the values

**Examples of valid syntaxes**

\[
\begin{align*}
ds \,:= \, ds_2 \text{ in } \{1,4,6\} & \quad \text{as usual, here the braces denote a set literal (it contains the values 1, 4 and 6)} \\
ds \,:= \, ds_3 \text{ in mySet} \\
ds \,:= \, ds_3 \text{ in myValueDomain}
\end{align*}
\]

**Semantics for scalar operations**

- **The in operator** returns \( \text{TRUE} \) if \( \text{op} \) belongs to the collection, \( \text{FALSE} \) otherwise.
- **The not_in operator** returns \( \text{FALSE} \) if \( \text{op} \) belongs to the collection, \( \text{TRUE} \) otherwise.

For example:

\[
\begin{align*}
1 & \text{ in } \{1,2,3\} & \text{returns } & \text{TRUE} \\
"a" & \text{ in } \{"c","ab","bb","bc"\} & \text{returns } & \text{FALSE} \\
"b" & \text{ not_in } \{"b","hello","c"\} & \text{returns } & \text{FALSE} \\
"b" & \text{ not_in } \{"a","hello","c"\} & \text{returns } & \text{TRUE}
\end{align*}
\]

**Input parameters type**

- **op**: dataset \{measure<scalar> \_\_ \}
- \_\_ component<scalar>
- \_\_ scalar
- **collection**: set<scalar> \_ name<value_domain>

**Result type**

- **result**: dataset \{ measure<boolean> bool_var \}
- \_\_ component<boolean>
- \_\_ boolean

**Additional constraints**

- The operand must be of a basic scalar data type compatible with the basic scalar type of the collection.

**Behaviour**

**Semantics**

- The in operator evaluates to \( \text{TRUE} \) if the operand is an element of the specified collection and \( \text{FALSE} \) otherwise, the not_in the opposite.
- The operator has the typical behaviour of the “Operators changing the data type” (see the section “Typical behaviours of the ML Operators”).
- The collection can be either a set of values defined in line or a name that references an externally defined Value Domain or Set.

**Examples**

Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>DS_1</th>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>2012</td>
<td>BS</td>
<td>0</td>
</tr>
</tbody>
</table>
Example 1:

\[
DS_r := DS_1 \text{ in } \{0, 3, 6, 12\}
\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>BS</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>GZ</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>SQ</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>MO</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>FJ</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>CQ</td>
<td>FALSE</td>
</tr>
</tbody>
</table>

Example 2 (on Components):

\[
DS_r := DS_1 \text{ [ calc Me}_2 := \text{ Me}_1 \text{ in } \{0, 3, 6, 12\} ]
\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>BS</td>
<td>0</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>GZ</td>
<td>4</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>SQ</td>
<td>9</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>MO</td>
<td>6</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>FJ</td>
<td>7</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>CQ</td>
<td>2</td>
<td>FALSE</td>
</tr>
</tbody>
</table>

Given the previous Data Set DS_1 and the following Value Domain named myGeoValueDomain (which has the basic scalar type \textit{string}):

\[
\begin{array}{|c|c|}
\hline
\text{Code} & \text{Meaning} \\
\hline
AF & Afghanistan \\
BS & Bahamas \\
FJ & Fiji \\
GA & Gabon \\
KH & Cambodia \\
MO & Macao \\
PK & Pakistan \\
QA & Qatar \\
\hline
\end{array}
\]
Example 3 (on external Value Domain):

\[ DS_r := DS_1 \#Id_2 \text{ in myGeoValueDomain} \]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>BS</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>GZ</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>SQ</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>MO</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>FJ</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>CQ</td>
<td>FALSE</td>
</tr>
</tbody>
</table>

**match_characters**

**Syntax**

\[
\text{match_characters ( op, pattern )}
\]

**Input parameters**

- **op** the dataset to be checked
- **pattern** the regular expression to check the Data Set or the Component against

**Examples of valid syntaxes**

\[
\text{match_characters(ds1, "[abc]\+d\d")}
\]

\[
ds1 \begin{array}{l}
\text{calc m1 := match_characters(ds1, "[abc]\+d\d")}
\end{array}
\]

**Semantics for scalar operations**

**match_characters** returns TRUE if op matches the regular expression regexp, FALSE otherwise. The string regexp is an Extended Regular Expression as described in the POSIX standard. Different implementations of VTL may implement different versions of the POSIX standard therefore it is possible that **match_characters** may behave in slightly different ways.

**Input parameters type**

- **op** dataset {measure<string> _}
  - component<string>
  - string
- **pattern** string | component<string>

**Result type**

- dataset { measure<boolean> bool_var }
  - component<boolean>
  - boolean

**Additional constraints**

If op is a Data Set then it has exactly one measure.
pattern is a POSIX regular expression.

**Behaviour**
The operator has the typical behaviour of the “Operators changing the data type” (see the section “Typical behaviours of the ML Operators”).

**Examples**
Given the following Dataset DS_1:

<table>
<thead>
<tr>
<th></th>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>G</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td>AX123</td>
<td></td>
</tr>
<tr>
<td>R</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td>AX2J5</td>
<td></td>
</tr>
</tbody>
</table>

DS_r := (ds1, "[:alpha:][2][:digit:][3]") results in:

<table>
<thead>
<tr>
<th></th>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>G</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td>TRUE</td>
<td></td>
</tr>
<tr>
<td>R</td>
<td>Total</td>
<td>Percentage</td>
<td>Total</td>
<td>FALSE</td>
<td></td>
</tr>
</tbody>
</table>

**Isnull:**

**isnull**

**Syntax**

isnull ( op )

**Input parameters**

operand mandatory the operand

**Examples of valid syntaxes**

isnull(DS_1)

**Semantics for scalar operations**
The operator returns TRUE if the value of the operand is NULL, FALSE otherwise.

**Examples**

isnull("Hello") gives: FALSE

isnull(NULL) gives: TRUE

**Input parameters type**

op :: dataset {measure<scalar> .}  
    | component<scalar>
    | scalar

**Result type**

result :: dataset { measure<boolean> bool_var }  
    | component<boolean>
    | boolean

**Additional constraints**

If op is a Data Set then it has exactly one measure.

**Behaviour**
The operator has the typical behaviour of the “Operators changing the data type” (see the section “Typical behaviours of the ML Operators”).

Examples

Given the operand Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>11094850</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>11123034</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>417546</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>5401267</td>
</tr>
<tr>
<td>2012</td>
<td>N</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
</tr>
</tbody>
</table>

Example 1: DS_r := isnull(DS_1) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>N</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Example 2 (on Components): DS_r := DS_1[ calc Me_2 := isnull(Me_1) ] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>11094850</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>11123034</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>417546</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>5401267</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>N</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Exists in : exists_in

Syntax

exists_in ( op1, op2 { , retain } )

retain ::= true | false | all

Input parameters
Examples of valid syntaxes
- `exists_in ( DS_1, DS_2, true )`
- `exists_in ( DS_1, DS_2 )`
- `exists_in ( DS_1, DS_2, all )`

Semantics for scalar operations
This operator cannot be applied to scalar values.

Input parameters type
- `op1`, `op2` :: dataset

Result type
- `result` :: dataset { measure<boolean> bool_var }

Additional constraints
- `op1` has at least all the identifier components of `op2` or `op2` has at least all the identifier components of `op1`.

Behaviour
The operator takes under consideration the common Identifiers of `op1` and `op2` and checks if the combinations of values of these Identifiers which are in `op1` also exist in `op2`.
- The result has the same Identifiers as `op1` and a `boolean` Measure `bool_var` whose value, for each Data Point of `op1`, is `TRUE` if the combination of values of the common Identifier Components in `op1` is found in a Data Point of `op2`, `FALSE` otherwise.
- If `retain` is `all` then both the Data Points having `bool_var` = `TRUE` and `bool_var` = `FALSE` are returned. If `retain` is `true` then only the data points with `bool_var` = `TRUE` are returned. If `retain` is `false` then only the Data Points with `bool_var` = `FALSE` are returned. If the `retain` parameter is omitted, the default is `all`.
- The operator has the typical behaviour of the "Operators changing the data type" (see the section "Typical behaviours of the ML Operators").

Examples
Given the operand Data Sets `DS_1` and `DS_2`:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>11094850</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>11123034</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>46818219</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>417546</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>5401267</td>
</tr>
<tr>
<td>2012</td>
<td>W</td>
<td>Total</td>
<td>Total</td>
<td>7954662</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>0.023</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>M</td>
<td>0.286</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>0.064</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>M</td>
<td>0.043</td>
</tr>
</tbody>
</table>
Example 1: \( DS_r := \text{exists\_in} (DS_1, DS_2, \text{all}) \)  
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>W</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Example 2: \( DS_r := \text{exists\_in} (DS_1, DS_2, \text{true}) \)  
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
<tr>
<td>2012</td>
<td>W</td>
<td>Total</td>
<td>Total</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Example 3: \( DS_r := \text{exists\_in} (DS_1, DS_2, \text{false}) \)  
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>bool_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>FALSE</td>
</tr>
</tbody>
</table>
Logical conjunction:  \textbf{and}

\textit{Syntax}
\begin{eqnarray*}
\text{op1} \text{ and } \text{op2}
\end{eqnarray*}

\textit{Input parameters}
\begin{itemize}
\item op1 \quad \text{the first operand}
\item op2 \quad \text{the second operand}
\end{itemize}

\textit{Examples of valid syntaxes}
\begin{itemize}
\item DS_1 and DS_2
\end{itemize}

\textit{Semantics for scalar operations}
The \textbf{and} operator returns TRUE if both operands are TRUE, otherwise FALSE. The two operands must be of \textit{boolean} type.

For example:
\begin{itemize}
\item FALSE and FALSE \quad \text{gives} \quad FALSE
\item FALSE and TRUE \quad \text{gives} \quad FALSE
\item FALSE and NULL \quad \text{gives} \quad FALSE
\item TRUE and FALSE \quad \text{gives} \quad FALSE
\item TRUE and TRUE \quad \text{gives} \quad TRUE
\item TRUE and NULL \quad \text{gives} \quad NULL
\item NULL and NULL \quad \text{gives} \quad NULL
\end{itemize}

\textit{Input parameters type}
\begin{itemize}
\item op1, op2 :: dataset \{ \text{measure<boolean> \_} \\
\item | \text{component<boolean>}
\item | boolean
\end{itemize}

\textit{Result type}
\begin{itemize}
\item result :: dataset \{ \text{measure<boolean> \_} \\
\item | \text{component<boolean>}
\item | boolean
\end{itemize}

\textit{Additional constraints}
None.

\textit{Behaviour}
The operator has the typical behaviour of the “Behaviour of Boolean operators” (see the section “Typical behaviours of the ML Operators”).

\textit{Examples}
Given the operand Data Sets DS_1 and DS_2:

\begin{tabular}{|c|c|c|c|c|}
\hline
\textbf{DS_1} & \\
\hline
\text{Id}_1 & \text{Id}_2 & \text{Id}_3 & \text{Id}_4 & \text{Me}_1 \\
\hline
M & 15 & B & 2013 & TRUE \\
M & 64 & B & 2013 & FALSE \\
M & 65 & B & 2013 & TRUE \\
F & 15 & U & 2013 & FALSE \\
\hline
\end{tabular}
Example 1: \( DS_r := DS_1 \) and \( DS_2 \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>15</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
</tbody>
</table>

Example 2 (on Components): \( DS_r := DS_1 \) \[\text{calc} Me_2 := Me_1 \) and true \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
<td>FALSE</td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>15</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Logical disjunction: \( \text{or} \)

Syntax

\[ \text{op1 or op2} \]

Input parameters

- \( \text{op1} \) the first operand
- \( \text{op2} \) the second operand

Examples of valid syntaxes

- \( DS_1 \) or \( DS_2 \)
Semantics for scalar operations

The or operator returns TRUE if at least one of the operands is TRUE, otherwise FALSE. The two operands must be of boolean type.

For example:

- FALSE or FALSE gives FALSE
- FALSE or TRUE gives TRUE
- FALSE or NULL gives NULL
- TRUE or FALSE gives TRUE
- TRUE or TRUE gives TRUE
- TRUE or NULL gives TRUE
- NULL or NULL gives NULL

Input parameters type

\[
\text{op1, op2 :: dataset \{measure<boolean> \_\}
\]

<table>
<thead>
<tr>
<th>component&lt;boolean&gt;</th>
</tr>
</thead>
<tbody>
<tr>
<td>boolean</td>
</tr>
</tbody>
</table>

Result type

\[
\text{result :: dataset \{ measure<boolean> \_\}
\]

<table>
<thead>
<tr>
<th>component&lt;boolean&gt;</th>
</tr>
</thead>
<tbody>
<tr>
<td>boolean</td>
</tr>
</tbody>
</table>

Additional constraints

None.

Behaviour

The operator has the typical behaviour of the “Behaviour of Boolean operators” (see the section “Typical behaviours of the ML Operators”).

Examples

Given the operand Data Sets DS_1 and DS_2:

### DS_1

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>15</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

### DS_2

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>15</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
</tbody>
</table>

Example 1: \( DS_r := DS_1 \ or \ DS_2 \) results in:
Example 2 (on Components):

\[
\text{DS}_r := \text{DS}_1 \left[ \text{calc } \text{Me}_2 := \text{Me}_1 \text{ or true} \right]
\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>15</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

**Exclusive disjunction:** xor

**Syntax**

\[
\text{op1} \text{ xor } \text{op2}
\]

**Input parameters**

- \text{op1} the first operand
- \text{op2} the second operand

**Examples of valid syntaxes**

- \text{DS}_1 \text{ xor } \text{DS}_2

**Semantics for scalar operations**

The xor operator returns TRUE if only one of the operand is TRUE (but not both), FALSE otherwise. The two operands must be of boolean type.

For example:

- FALSE xor FALSE gives FALSE
- FALSE xor TRUE gives TRUE
- FALSE xor NULL gives NULL
- TRUE xor FALSE gives TRUE
- TRUE xor TRUE gives FALSE
- TRUE xor NULL gives NULL
- NULL xor NULL gives NULL

**Input parameters type**

- \text{op1}, \text{op2} ::

  dataset {measure<boolean> _ }

  | component<boolean>
  |

  | boolean
Result type

result:: dataset { measure<boolean> _ }  
| component<boolean>  
| boolean

Additional constraints

None.

Behaviour

The operator has the typical behaviour of the “Behaviour of Boolean operators” (see the section “Typical behaviours of the ML Operators”).

Examples

Given the operand Data Sets DS_1 and DS_2:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>15</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
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<td>TRUE</td>
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<td>15</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
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<tr>
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<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
</tbody>
</table>

Example 1:  

DS_r:= DS_1 xor DS_2  

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>15</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
</tr>
</tbody>
</table>

Example 2 (on Components):  

DS_r:= DS_1 [ calc Me_2:= Me_1 xor true ]  

results in:
Logical negation: \texttt{not}

\textbf{Syntax}

\begin{verbatim}
\texttt{not op}
\end{verbatim}

\textbf{Input parameters}

\begin{verbatim}
op \quad \text{the operand}
\end{verbatim}

\textbf{Examples of valid syntaxes}

\begin{verbatim}
not DS_1
\end{verbatim}

\textbf{Semantics for scalar operations}

The \texttt{not} operator returns \texttt{TRUE} if \texttt{op} is \texttt{FALSE}, otherwise \texttt{TRUE}. The input operand must be of \texttt{boolean} type. For example:

\begin{verbatim}
not FALSE gives TRUE
not TRUE gives FALSE
not NULL gives NULL
\end{verbatim}

\textbf{Input parameters type}

\begin{verbatim}
op ::
| \quad \text{dataset \{measure\texttt{<boolean>} \}}
| \quad | \text{component\texttt{<boolean>}}
| \quad | \text{boolean}
\end{verbatim}

\textbf{Result type}

\begin{verbatim}
result ::
| \quad \text{dataset \{measure\texttt{<boolean>} \}}
| \quad | \text{component\texttt{<boolean>}}
| \quad | \text{boolean}
\end{verbatim}

\textbf{Additional constraints}

None.

\textbf{Behaviour}

The operator has the typical behaviour of the “Behaviour of Boolean operators” (see the section “Typical behaviours of the ML Operators”).

\textbf{Examples}

Given the operand Data Set DS_1:
Example 1: \( DS_r := \neg DS_1 \) results in:

<table>
<thead>
<tr>
<th>DS_r</th>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>M</td>
<td>65</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
<td></td>
</tr>
<tr>
<td>F</td>
<td>15</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
<td></td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
<td></td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
<td></td>
</tr>
</tbody>
</table>

Example 2 (on Components): \( DS_r := DS_1 \ [\text{calc} \ Me_2 := \neg Me_1 \] results in:

<table>
<thead>
<tr>
<th>DS_r</th>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>M</td>
<td>15</td>
<td>B</td>
<td>2013</td>
<td>TRUE</td>
<td>FALSE</td>
<td></td>
</tr>
<tr>
<td>M</td>
<td>64</td>
<td>B</td>
<td>2013</td>
<td>FALSE</td>
<td>TRUE</td>
<td></td>
</tr>
<tr>
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<td>65</td>
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<td>2013</td>
<td>TRUE</td>
<td>FALSE</td>
<td></td>
</tr>
<tr>
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<td>15</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
<td>TRUE</td>
<td></td>
</tr>
<tr>
<td>F</td>
<td>64</td>
<td>U</td>
<td>2013</td>
<td>FALSE</td>
<td>TRUE</td>
<td></td>
</tr>
<tr>
<td>F</td>
<td>65</td>
<td>U</td>
<td>2013</td>
<td>TRUE</td>
<td>FALSE</td>
<td></td>
</tr>
</tbody>
</table>
This chapter describes the **time** operators, which are the operators dealing with **time**, **date** and **time_period** basic scalar types. The general aspects of the behaviour of these operators is described in the section “Behaviour of the Time Operators”.

The **time** data type is the most general type and denotes a generic time interval, having start and end points in time and therefore a duration, which is the time intervening between the start and end points. The **date** data type denotes a generic time instant (a point in time), which is a time interval with zero duration. The **time_period** data type denotes a regular time interval whose regular duration is explicitly represented inside each **time_period** value and is named **period_indicator**. In some sense, we say that **date** and **time_period** are special cases of **time**, the former with coinciding extremes and zero duration and the latter with regular duration. The **time** data type is overarching in the sense that it comprises **date** and **time_period**. Finally, **duration** data type represents a generic time span, independently of any specific start and end date.

The time, date and time period formats used here are explained in the User Manual in the section “External representations and literals used in the VTL Manuals”.

The period indicator **P id** of the **duration** type and its possible values are:

- **D** Day
- **W** Week
- **M** Month
- **Q** Quarter
- **S** Semester
- **A** Year

As already said, these representation are not prescribed by VTL and are not part of the VTL standard, each VTL system can personalize the representation of time, date, **time_period** and duration as desired. The formats shown above are only the ones used in the examples.

For a fully-detailed explanation, please refer to the User Manual.

### Period indicator: **period_indicator**

The operator **period_indicator** extracts the period indicator from a **time_period** value.

**Syntax**

```
period_indicator ( { op } )
```

**Input parameters**

- **op** the operand

**Examples of valid syntaxes**

- `period_indicator ( ds_1 )` (if used in a clause the operand **op** can be omitted)

**Semantics for scalar operations**

**period_indicator** returns the period indicator of a **time_period** value. The period indicator is the part of the **time_period** value which denotes the duration of the time period (e.g. day, week, month ...).

**Input parameters type**

```
op ::
    | dataset { identifier <time_period>*, identifier* }
    | component<time_period>
    | time_period
```

**Result type**

```
result ::
    | dataset { measure<duration> duration_var }
    | component <duration>
    | duration
```
**Additional constraints**

If \( op \) is a Data Set then it has exactly an Identifier of type `time_period` and may have other Identifiers. If the operator is used in a clause and \( op \) is omitted, then the Data Set to which the clause is applied has exactly an Identifier of type `time_period` and may have other Identifiers.

**Behaviour**

The operator extracts the period indicator part of the `time_period` value. The period indicator is computed for each Data Point. When the operator is used in a clause, it extracts the period indicator from the `time_period` value the Data Set to which the clause is applied.

The operator returns a Data Set with the same Identifiers of \( op \) and one Measure of type `duration` named `duration_var`. As for all the Variables, a proper Value Domain must be defined to contain the possible values of the period indicator and `duration_var`. The values used in the examples are listed at the beginning of this chapter "VTL-ML Time operators".

**Examples**

Given the Data Set \( DS_1 \):

\[
\begin{array}{cccc}
\text{Id}_1 & \text{Id}_2 & \text{Id}_3 & \text{Me}_1 \\
A & 1 & 2010 & 10 \\
A & 1 & 2013Q1 & 50 \\
\end{array}
\]

**Example 1:** \( DS_r := \text{period_indicator} ( DS_1 ) \) results in:

\[
\begin{array}{cccc}
\text{Id}_1 & \text{Id}_2 & \text{Id}_3 & \text{duration}_\text{var} \\
A & 1 & 2010 & A \\
A & 1 & 2013Q1 & Q \\
\end{array}
\]

**Example 2 (on component):** \( DS_r := DS_1 [ \text{filter period_indicator ( \text{Id}_3 ) = "A" } ] \) results in:

\[
\begin{array}{cccc}
\text{Id}_1 & \text{Id}_2 & \text{Id}_3 & \text{Me}_1 \\
A & 1 & 2010 & 10 \\
\end{array}
\]

**Fill time series**: \( \text{fill\_time\_series} \)

**Syntax**

\[
\text{fill\_time\_series} ( \text{op } \{ , \text{limitsMethod} \} )
\]

\[
\text{limitsMethod} ::= \text{single} | \text{all}
\]

**Input parameters**

\( \text{op} \) the operand

\( \text{limitsMethod} \) method for determining the limits of the time interval to be filled (default: \text{all})

**Examples of valid syntaxes**

\( \text{fill\_time\_series} ( \text{ds} ) \)

\( \text{fill\_time\_series} ( \text{ds, all} ) \)
Semantics for scalar operations

The fill_time_series operator does not perform scalar operations.

Input parameters type:

\[ \text{op} :: \text{dataset} \{ \text{identifier} < \text{time} > \_ , \text{identifier} \_ * \} \]

Result type:

\[ \text{result} :: \text{dataset} \{ \text{identifier} < \text{time} > \_ , \text{identifier} \_ * \} \]

Additional constraints

The operand \( \text{op} \) has an Identifier of type \text{time}, \text{date} or \text{time_period} and may have other Identifiers.

Behaviour

This operator can be applied only on Data Sets of time series and returns a Data Set of time series. The operator fills the possibly missing Data Points of all the time series belonging to the operand \( \text{op} \) within the time limits automatically determined by applying the limit_method.

If limitsMethod is \text{all}, the time limits are determined with reference to all the time_series of the Data Set: the limits are the minimum and the maximum values of the reference time Identifier Component of the Data Set.

If limitsMethod is \text{single}, the time limits are determined with reference to each single time_series of the Data Set: the limits are the minimum and the maximum values of the reference time Identifier Component of the time series.

The expected Data Points are determined, for each time series, by considering the limits above and the period (frequency) of the time series: all the Identifiers are kept unchanged except the reference time Identifier, which is increased of one period at a time (e.g. day, week, month, quarter, year) from the lower to the upper time limit.

For each increase, an expected Data Point is identified.

If this expected Data Points is missing, it is added to the Data Set. For the added Data Points, Measures and Attributes assume the NULL value.

The output Data Set has the same Identifier, Measure and Attribute Components as the operand Data Set. The output Data Set contains the same time series as the operand, because the time series Identifiers (all the Identifiers except the reference time Identifier) are not changed.

As mentioned in the section "Behaviour of the Time Operators", the operator is assumed to know which is the reference time Identifier as well as the period of each time series.

Examples

As described in the User Manual, the \text{time} data type is the intervening time between two time points and using the ISO 8601 standard it can be expressed through a start date and an end date separated by a slash at any precision. In the examples relevant to the \text{time} data type the precision is set at the level of month and the time format YYYY-MM/MM/YYYY is used.

Given the Data Set \( \text{DS}_1 \), which contains annual time series, where \( \text{Id}_2 \) is the reference time Identifier of \text{time} type:

<table>
<thead>
<tr>
<th>\text{Id}_1</th>
<th>\text{Id}_2</th>
<th>\text{Me}_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-01/2010-12</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2012-01/2012-12</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013-01/2013-12</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2011-01/2011-12</td>
<td>&quot;hi, hello! &quot;</td>
</tr>
<tr>
<td>B</td>
<td>2012-01/2012-12</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2014-01/2014-12</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>

Example 1: \( \text{DS}_r := \text{fill_time_series} \left( \text{DS}_1, \text{single} \right) \) results in:
Example 2: \( DS_r := \text{fill\_time\_series} \ ( DS_1, \text{all} ) \) results in:

\[
\begin{array}{|c|c|c|}
\hline
\text{Id}_1 & \text{Me}_1 & \text{Id}_2 \\
\hline
A & "hello world" & 2010-01/2010-12 \\
A & NULL & 2011-01/2011-12 \\
A & "say hello" & 2012-01/2012-12 \\
A & "he" & 2013-01/2013-12 \\
B & "hi, hello! " & 2011-01/2011-12 \\
B & "hi" & 2012-01/2012-12 \\
B & NULL & 2013-01/2013-12 \\
B & "hello!" & 2014-01/2014-12 \\
\hline
\end{array}
\]

Example 3: \( DS_r := \text{fill\_time\_series} \ ( DS_2, \text{single} ) \) results in:

\[
\begin{array}{|c|c|c|}
\hline
\text{Id}_1 & \text{Me}_1 & \text{Id}_2 \\
\hline
A & "hello world" & 2010-01/2010-12 \\
A & NULL & 2011-01/2011-12 \\
A & "say hello" & 2012-01/2012-12 \\
A & "he" & 2013-01/2013-12 \\
B & "hi, hello! " & 2011-01/2011-12 \\
B & "hi" & 2012-01/2012-12 \\
B & NULL & 2013-01/2013-12 \\
B & "hello!" & 2014-01/2014-12 \\
\hline
\end{array}
\]

Given the Data Set \( DS_2 \), which contains \textit{annual} time series, where \text{Id}_2\ is the reference time \text{Identifier of date type} and conventionally each period is identified by its last day:

\[
\begin{array}{|c|c|c|}
\hline
\text{Id}_1 & \text{Me}_1 & \text{Id}_2 \\
\hline
A & "hello world" & 2010-12-31 \\
A & "say hello" & 2012-12-31 \\
A & "he" & 2013-12-31 \\
B & "hi, hello! " & 2011-12-31 \\
B & "hi" & 2012-12-31 \\
B & "hello!" & 2014-12-31 \\
\hline
\end{array}
\]
Given the Data Set DS_3, which contains annual time series, where Id_2 is the reference time Identifier of time_period type:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>&quot;hi, hello!&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2012</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2013</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2014</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>

Example 5: DS_r := fill_time_series ( DS_3, single ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>&quot;hi, hello!&quot;</td>
</tr>
</tbody>
</table>

Example 4: DS_r := fill_time_series ( DS_2, all ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2014</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>&quot;hi, hello!&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2012</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2013</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2014</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>
Example 6: \[ \text{DS}_r := \text{fill\_time\_series} \left( \text{DS}_3, \text{all} \right) \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2014</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2010</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>&quot;hi, hello! &quot;</td>
</tr>
<tr>
<td>B</td>
<td>2012</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2013</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2014</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>

Example 7: \[ \text{DS}_r := \text{fill\_time\_series} \left( \text{DS}_4, \text{single} \right) \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2010Q1</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2010Q2</td>
<td>&quot;hi, hello! &quot;</td>
</tr>
<tr>
<td>A</td>
<td>2010Q4</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2011Q2</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>

Given the Data Set \( \text{DS}_4 \), which contains both \textit{quarterly} and \textit{annual} time series relevant to the same phenomenon "A", where \( \text{Id}_2 \) is the reference time Identifier of \textit{time\_period} type,
Example 8: \[ DS_r := \text{fill\_time\_series} \left( DS_4, \text{all} \right) \]
results in:

\[
\begin{array}{|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 \\
\hline
\text{A} & 2010 & \text{"hello world"} \\
\text{A} & 2011 & \text{NULL} \\
\text{A} & 2012 & \text{"say hello"} \\
\text{A} & 2010Q1 & \text{"he"} \\
\text{A} & 2010Q2 & \text{"hi, hello!"} \\
\text{A} & 2010Q3 & \text{NULL} \\
\text{A} & 2010Q4 & \text{"hi"} \\
\text{A} & 2011Q1 & \text{NULL} \\
\text{A} & 2011Q2 & \text{"hello!"} \\
\text{A} & 2011Q3 & \text{NULL} \\
\text{A} & 2011Q4 & \text{NULL} \\
\text{A} & 2012Q1 & \text{NULL} \\
\text{A} & 2012Q2 & \text{NULL} \\
\text{A} & 2012Q3 & \text{NULL} \\
\text{A} & 2012Q4 & \text{NULL} \\
\hline
\end{array}
\]
Flow to stock :  flow_to_stock

Syntax

\[
\text{flow_to_stock} \ (\ op )
\]

Input Parameters

\begin{itemize}
\item \text{op} \quad \text{the operand}
\end{itemize}

Examples of valid syntaxes

\begin{itemize}
\item \text{flow_to_stock} \ (\ ds_1 )
\end{itemize}

Semantics for scalar operations

This operator does not perform scalar operations.

Input parameters type:

\begin{itemize}
\item \text{op} :: \quad \text{dataset} \{ \text{identifier} \ < \text{time} > \_\_ , \text{identifier} \ _* , \text{measure} < \text{number} > \ _+ \}
\end{itemize}

Result type:

\begin{itemize}
\item \text{result} :: \quad \text{dataset} \{ \text{identifier} \ < \text{time} > \_\_ , \text{identifier} \ _* , \text{measure} < \text{number} > \ _+ \}
\end{itemize}

Additional constraints

The operand dataset has an Identifier of type \text{time}, \text{date} or \text{time_period} and may have other Identifiers.

Behaviour

The statistical data that describe the “state” of a phenomenon on a given moment (e.g. resident population on a given moment) are often referred to as “stock data”. On the contrary, the statistical data that describe “events” which can happen continuously (e.g. changes in the resident population, such as births, deaths, immigration, emigration), are often referred to as “flow data”. This operator takes in input a Data Set which are interpreted as flows and calculates the change of the corresponding stock since the beginning of each time series by summing the relevant flows. In other words, the operator perform the cumulative sum from the first Data Point of each time series to each other following Data Point of the same time series. The \text{flow_to_stock} operator can be applied only on Data Sets of time series and returns a Data Set of time series. The result Data Set has the same Identifier, Measure and Attribute Components as the operand Data Set and contains the same time series as the operand, because the time series Identifiers (all the Identifiers except the reference time Identifier) are not changed. As mentioned in the section “Behaviour of the Time Operators”, the operator is assumed to know which is the \text{time} Identifier as well as the \text{period} of each time series.

Examples

As described in the User Manual, the \text{time} data type is the intervening time between two time points and using the ISO 8601 standard it can be expressed through a start date and an end date separated by a slash at any precision. In the examples relevant to the \text{time} data type the precision is set at the level of month and the time format YYYY-MM/YYY-MM is used.

Given the Data Set \text{DS_1}, which contains \text{annual} time series, where \text{Id_2} is the reference time Identifier of \text{time} type:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-01/2010-12</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011-01/2011-12</td>
<td>5</td>
</tr>
</tbody>
</table>
Example 1:  
\[ DS_r := \text{flow_to_stock} \left( DS_1 \right) \]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-01/2010-12</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011-01/2011-12</td>
<td>7</td>
</tr>
<tr>
<td>A</td>
<td>2012-01/2012-12</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>2013-01/2013-12</td>
<td>13</td>
</tr>
<tr>
<td>B</td>
<td>2010-01/2010-12</td>
<td>4</td>
</tr>
<tr>
<td>B</td>
<td>2011-01/2011-12</td>
<td>-4</td>
</tr>
<tr>
<td>B</td>
<td>2012-01/2012-12</td>
<td>-4</td>
</tr>
<tr>
<td>B</td>
<td>2013-01/2013-12</td>
<td>2</td>
</tr>
</tbody>
</table>

Example 2:  
\[ DS_r := \text{flow_to_stock} \left( DS_2 \right) \]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-12-31</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011-12-31</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>2012-12-31</td>
<td>-3</td>
</tr>
<tr>
<td>A</td>
<td>2013-12-31</td>
<td>9</td>
</tr>
<tr>
<td>B</td>
<td>2010-12-31</td>
<td>4</td>
</tr>
<tr>
<td>B</td>
<td>2011-12-31</td>
<td>-8</td>
</tr>
<tr>
<td>B</td>
<td>2012-12-31</td>
<td>0</td>
</tr>
<tr>
<td>B</td>
<td>2013-12-31</td>
<td>6</td>
</tr>
</tbody>
</table>

Given the Data Set \( DS_2 \), which contains *annual* time series, where \( Id_2 \) is the reference time Identifier of *date* type (conventionally each period is identified by its last day):

\( DS_r \)
<table>
<thead>
<tr>
<th>Year</th>
<th>Time Period</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>12-31-2011</td>
<td>-4</td>
</tr>
<tr>
<td>2012</td>
<td>12-31-2012</td>
<td>-4</td>
</tr>
<tr>
<td>2013</td>
<td>12-31-2013</td>
<td>2</td>
</tr>
</tbody>
</table>

Given the Data Set DS_3, which contains *annual* time series, where \( \text{Id}_2 \) is the reference time Identifier of *time_period* type:

<table>
<thead>
<tr>
<th>( \text{Id}_1 )</th>
<th>( \text{Id}_2 )</th>
<th>( \text{Me}_1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>-3</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>9</td>
</tr>
<tr>
<td>B</td>
<td>2010</td>
<td>4</td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>-8</td>
</tr>
<tr>
<td>B</td>
<td>2012</td>
<td>0</td>
</tr>
<tr>
<td>B</td>
<td>2013</td>
<td>6</td>
</tr>
</tbody>
</table>

**Example 3:**

\[ DS_r := \text{flow_to_stock} \left( DS_3 \right) \]

results in:

<table>
<thead>
<tr>
<th>( \text{Id}_1 )</th>
<th>( \text{Id}_2 )</th>
<th>( \text{Me}_1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>7</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>13</td>
</tr>
<tr>
<td>B</td>
<td>2010</td>
<td>4</td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>-4</td>
</tr>
<tr>
<td>B</td>
<td>2012</td>
<td>-4</td>
</tr>
<tr>
<td>B</td>
<td>2013</td>
<td>2</td>
</tr>
</tbody>
</table>

Given the Data Set DS_4, which contains both *quarterly* and *annual* time series relevant to the same phenomenon “A”, where \( \text{Id}_2 \) is the reference time Identifier of *time_period* type:

<table>
<thead>
<tr>
<th>( \text{Id}_1 )</th>
<th>( \text{Id}_2 )</th>
<th>( \text{Me}_1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>7</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>13</td>
</tr>
<tr>
<td>A</td>
<td>2010Q1</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2010Q2</td>
<td>-3</td>
</tr>
<tr>
<td>A</td>
<td>2010Q3</td>
<td>7</td>
</tr>
</tbody>
</table>
Example 4: \( DS_r := \text{flow\_to\_stock} ( DS_3 ) \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>13</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>26</td>
</tr>
<tr>
<td>A</td>
<td>2010Q1</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2010Q2</td>
<td>-1</td>
</tr>
<tr>
<td>A</td>
<td>2010Q3</td>
<td>6</td>
</tr>
<tr>
<td>A</td>
<td>2010Q4</td>
<td>2</td>
</tr>
</tbody>
</table>

Stock to flow: \( \text{stock\_to\_flow} \)

**Syntax**
\[
\text{stock\_to\_flow} \ ( \text{op} )
\]

**Input parameters**
- \( \text{op} \) the operand

**Examples of valid syntaxes**
- \( \text{stock\_to\_flow} \ ( \text{ds\_1} ) \)

**Semantics for scalar operations**
This operator does not perform scalar operations.

**Input parameters type:**
- \( \text{op} :: \) dataset \{ identifier < time \_, identifier \_*, measure<number> \_+ \}

**Result type:**
- \( \text{result} :: \) dataset \{ identifier < time \_, identifier \_*, measure<number> \_+ \}

**Additional constraints**
The operand dataset has an Identifier of type \( \text{time, date or time\_period} \) and may have other Identifiers.

**Behaviour**
The statistical data that describe the “state” of a phenomenon on a given moment (e.g. resident population on a given moment) are often referred to as “stock data”.
On the contrary, the statistical data that describe “events” which can happen continuously (e.g. changes in the resident population, such as births, deaths, immigration, emigration), are often referred to as “flow data”.
This operator takes in input a Data Set of time series which is interpreted as stock data and, for each time series, calculates the corresponding flow data by subtracting from the measure values of each regular period the corresponding measure values of the previous one.
The \( \text{stock\_to\_flow} \) operator can be applied only on Data Sets of time series and returns a Data Set of time series.
The result Data Set has the same Identifier, Measure and Attribute Components as the operand Data Set and contains the same time series as the operand, because the time series Identifiers (all the Identifiers except the reference time Identifier) are not changed.
The Attribute propagation rule is not applied.

As mentioned in the section "Behaviour of the Time Operators", the operator is assumed to know which is the time Identifier as well as the period of each time series.

Examples

As described in the User Manual, the time data type is the intervening time between two time points and using the ISO 8601 standard it can be expressed through a start date and an end date separated by a slash at any precision. In the examples relevant to the time data type the precision is set at the level of month and the time format YYYY-MM/YYYY-MM is used.

Given the Data Set DS_1, which contains annual time series, where ld_2 is the reference time Identifier of time type:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-01/2010-12</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011-01/2011-12</td>
<td>7</td>
</tr>
<tr>
<td>A</td>
<td>2012-01/2012-12</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>2013-01/2013-12</td>
<td>13</td>
</tr>
<tr>
<td>B</td>
<td>2010-01/2010-12</td>
<td>4</td>
</tr>
<tr>
<td>B</td>
<td>2011-01/2011-12</td>
<td>-4</td>
</tr>
<tr>
<td>B</td>
<td>2012-01/2012-12</td>
<td>-8</td>
</tr>
<tr>
<td>B</td>
<td>2013-01/2013-12</td>
<td>2</td>
</tr>
</tbody>
</table>

Example 1: DS_r := stock_to_flow ( DS_1 ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-01/2010-12</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011-01/2011-12</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>2012-01/2012-12</td>
<td>-3</td>
</tr>
<tr>
<td>A</td>
<td>2013-01/2013-12</td>
<td>9</td>
</tr>
<tr>
<td>B</td>
<td>2010-01/2010-12</td>
<td>4</td>
</tr>
<tr>
<td>B</td>
<td>2011-01/2011-12</td>
<td>-8</td>
</tr>
<tr>
<td>B</td>
<td>2012-01/2012-12</td>
<td>0</td>
</tr>
<tr>
<td>B</td>
<td>2013-01/2013-12</td>
<td>6</td>
</tr>
</tbody>
</table>

Given the Data Set DS_2, which contains annual time series, where ld_2 is the reference time Identifier of date type (conventionally each period is identified by its last day):

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-12-31</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011-12-31</td>
<td>7</td>
</tr>
<tr>
<td>A</td>
<td>2012-12-31</td>
<td>4</td>
</tr>
</tbody>
</table>
Example 2: \[ DS_r := \text{stock\_to\_flow} ( DS_2 ) \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-12-31</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011-12-31</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>2012-12-31</td>
<td>-3</td>
</tr>
<tr>
<td>A</td>
<td>2013-12-31</td>
<td>9</td>
</tr>
<tr>
<td>B</td>
<td>2010-12-31</td>
<td>4</td>
</tr>
<tr>
<td>B</td>
<td>2011-12-31</td>
<td>-8</td>
</tr>
<tr>
<td>B</td>
<td>2012-12-31</td>
<td>0</td>
</tr>
<tr>
<td>B</td>
<td>2013-12-31</td>
<td>6</td>
</tr>
</tbody>
</table>

Given the Data Set \( DS_3 \), which contains *annual* time series, where Id_2 is the reference time Identifier of *time_period* type:

Example 3: \[ DS_r := \text{stock\_to\_flow} ( DS_3 ) \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>7</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>13</td>
</tr>
<tr>
<td>B</td>
<td>2010</td>
<td>4</td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>-4</td>
</tr>
<tr>
<td>B</td>
<td>2012</td>
<td>-4</td>
</tr>
<tr>
<td>B</td>
<td>2013</td>
<td>2</td>
</tr>
</tbody>
</table>
Given the Data Set DS_4, which contains both quarterly and annual time series relevant to the same phenomenon “A”, where \textit{Id}_2 is the \textit{time} \textit{Identifier} of \textit{time} \textit{period} type:

\begin{table}[h]
\centering
\begin{tabular}{|c|c|c|}
\hline
\textbf{Id}_1 & \textbf{Id}_2 & \textbf{Me}_1 \\
\hline
A & 2010 & 2  \\
A & 2011 & 9  \\
A & 2012 & 13  \\
A & 2013 & 26  \\
A & 2010Q1 & 2  \\
A & 2010Q2 & -1  \\
A & 2010Q3 & 6  \\
A & 2010Q4 & 2  \\
\hline
\end{tabular}
\end{table}

\textbf{Example 4:} \quad DS_r := \text{stock\_to\_flow} ( DS_4 ) \quad \text{results in:}

\begin{table}[h]
\centering
\begin{tabular}{|c|c|c|}
\hline
\textbf{Id}_1 & \textbf{Id}_2 & \textbf{Me}_1 \\
\hline
A & 2010 & 2  \\
A & 2011 & 7  \\
A & 2012 & 4  \\
A & 2013 & 13  \\
A & 2010Q1 & 2  \\
A & 2010Q2 & -3  \\
A & 2010Q3 & 7  \\
A & 2010Q4 & -4  \\
\hline
\end{tabular}
\end{table}

\textbf{Time shift: \quad timeshift}

\textbf{Syntax}

\texttt{timeshift ( op, shiftNumber )}

\textbf{Input parameters}

- \textit{op} \quad the operand
- \textit{shiftNumber} \quad the number of periods to be shifted

\textbf{Examples of valid syntaxes}

- \texttt{timeshift ( DS_1, 2 )}
- \texttt{timeshift ( DS_1, 1 )}

\textbf{Semantics for scalar operations}

This operator does not perform scalar operations.
Input parameters type:

```plaintext
4804 4805 4806 4807
4808 4809 4810 4811 4812 4813 4814
4815 4816 4817 4818 4819 4820 4821 4822 4823 4824 4825 4826 4827 4828 4829 4830 4831 4832 4833 4834 4835 4836 4837 4838
```

```plaintext
Input parameters type:

```
4804 4805 4806 4807
4808 4809 4810 4811 4812 4813 4814
4815 4816 4817 4818 4819 4820 4821 4822 4823 4824 4825 4826 4827 4828 4829 4830 4831 4832 4833 4834 4835 4836 4837 4838
```

Result type:

```plaintext
4808 4809 4810 4811 4812 4813 4814
4815 4816 4817 4818 4819 4820 4821 4822 4823 4824 4825 4826 4827 4828 4829 4830 4831 4832 4833 4834 4835 4836 4837 4838
```

```plaintext
Input parameters type:

```
4804 4805 4806 4807
4808 4809 4810 4811 4812 4813 4814
4815 4816 4817 4818 4819 4820 4821 4822 4823 4824 4825 4826 4827 4828 4829 4830 4831 4832 4833 4834 4835 4836 4837 4838
```

Result type:

```plaintext
4808 4809 4810 4811 4812 4813 4814
4815 4816 4817 4818 4819 4820 4821 4822 4823 4824 4825 4826 4827 4828 4829 4830 4831 4832 4833 4834 4835 4836 4837 4838
```

```plaintext
Additional constraints

The operand dataset has an Identifier of type time, date or time_period and may have other Identifiers.
```

Behaviour

This operator takes in input a Data Set of time series and, for each time series of the Data Set, shifts the reference time Identifier of a number of periods (of the time series) equal to the shift_number parameter. If shift_number is negative, the shift is in the past, otherwise in the future. For example, if the period of the time series is month and shift_number is -1 the reference time Identifier is shifted of two months in the past.

The operator can be applied only on Data Sets of time series and returns a Data Set of time series.

The result Data Set has the same Identifier, Measure and Attribute Components as the operand Data Set and contains the same time series as the operand, because the time series Identifiers (all the Identifiers except the reference time Identifier) are not changed.

The Attribute propagation rule is not applied.

As mentioned in the section "Behaviour of the Time Operators", the operator is assumed to know which is the time Identifier as well as the period of each data point.

Examples

As described in the User Manual, the time data type is the intervening time between two time points and using the ISO 8601 standard it can be expressed through a start date and an end date separated by a slash at any precision. In the examples relevant to the time data type the precision is set at the level of month and the time format YYYY-MM/YYYY-MM is used.

Given the Data Set DS_1, which contains yearly time series, where Id_2 is the reference time Identifier of time type:

```
<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-01/2010-12</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2011-01/2011-12</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2012-01/2012-12</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013-01/2013-12</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2010-01/2010-12</td>
<td>&quot;hi, hello!&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2011-01/2011-12</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2012-01/2012-12</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2013-01/2013-12</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>
```

Example 1:  DS_r := timeshift ( DS_1 , -1 ) results in:

```
<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2009-01/2009-12</td>
<td>&quot;hello world&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010-01/2010-12</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2011-01/2011-12</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2012-01/2012-12</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2009-01/2009-12</td>
<td>&quot;hi, hello!&quot;</td>
</tr>
</tbody>
</table>
```
Given the Data Set DS_2, which contains annual time series, where Id_2 is the reference time Identifier of date type (conventionally each period is identified by its last day):

<table>
<thead>
<tr>
<th>Id</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010-12-31</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2011-12-31</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2012-12-31</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013-12-31</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2010-12-31</td>
<td>&quot;hi, hello!&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2011-12-31</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2012-12-31</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2013-12-31</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>

Example 2: DS_r := timeshift ( DS_2 , 2 ) results in:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2012-12-31</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013-12-31</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2014-12-31</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2015-12-31</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2012-12-31</td>
<td>&quot;hi, hello!&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2013-12-31</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2014-12-31</td>
<td>NULL</td>
</tr>
<tr>
<td>B</td>
<td>2015-12-31</td>
<td>&quot;hello!&quot;</td>
</tr>
</tbody>
</table>

Given the Data Set DS_3, which contains annual time series, where Id_2 is the reference time Identifier of time_period type:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>&quot;hello world&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>&quot;say hello&quot;</td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>&quot;he&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2010</td>
<td>&quot;hi, hello!&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>&quot;hi&quot;</td>
</tr>
<tr>
<td>B</td>
<td>2012</td>
<td>NULL</td>
</tr>
</tbody>
</table>
Example 3: \[ DS_r := \text{timeshift} \left( DS_3, 1 \right) \]
results in:

<table>
<thead>
<tr>
<th>( DS_r )</th>
<th>( \text{id}_1 )</th>
<th>( \text{id}_2 )</th>
<th>( \text{Me}_1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2011</td>
<td>&quot;hello world&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>NULL</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>&quot;say hello&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2014</td>
<td>&quot;he&quot;</td>
<td></td>
</tr>
<tr>
<td>B</td>
<td>2011</td>
<td>&quot;hi, hello!&quot;</td>
<td></td>
</tr>
<tr>
<td>B</td>
<td>2012</td>
<td>&quot;hi&quot;</td>
<td></td>
</tr>
<tr>
<td>B</td>
<td>2013</td>
<td>NULL</td>
<td></td>
</tr>
<tr>
<td>B</td>
<td>2014</td>
<td>&quot;hello!&quot;</td>
<td></td>
</tr>
</tbody>
</table>

Given the Data Set \( DS_4 \), which contains both quarterly and annual time series relevant to the same phenomenon "A", where \( \text{id}_2 \) is the reference time Identifier of \( \text{time_period} \) type:

<table>
<thead>
<tr>
<th>( DS_4 )</th>
<th>( \text{id}_1 )</th>
<th>( \text{id}_2 )</th>
<th>( \text{Me}_1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2010</td>
<td>&quot;hello world&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>NULL</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>&quot;say hello&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2013</td>
<td>&quot;he&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010Q1</td>
<td>&quot;hi, hello!&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010Q2</td>
<td>&quot;hi&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010Q3</td>
<td>NULL</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010Q4</td>
<td>&quot;hello!&quot;</td>
<td></td>
</tr>
</tbody>
</table>

Example 4: \[ DS_r := \text{time_shift} \left( DS_3, -1 \right) \]
results in:

<table>
<thead>
<tr>
<th>( DS_r )</th>
<th>( \text{id}_1 )</th>
<th>( \text{id}_2 )</th>
<th>( \text{Me}_1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>2009</td>
<td>&quot;hello world&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010</td>
<td>NULL</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2011</td>
<td>&quot;say hello&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2012</td>
<td>&quot;he&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2009Q4</td>
<td>&quot;hi, hello!&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010Q1</td>
<td>&quot;hi&quot;</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010Q2</td>
<td>NULL</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>2010Q3</td>
<td>&quot;hello!&quot;</td>
<td></td>
</tr>
</tbody>
</table>
Time aggregation: \texttt{time\_agg}

The operator \texttt{time\_agg} converts \texttt{time}, \texttt{date} and \texttt{time\_period} values from a smaller to a larger duration.

Syntax

\begin{verbatim}
\texttt{time\_agg ( periodIndTo \{ , periodIndFrom \} \{ , op \} \{ , first | last \})}
\end{verbatim}

Input parameters

- \texttt{op}: the scalar value, the Component or the Data Set to be converted. If not specified, then time\_agg is used in combination within an aggregation operator.
- \texttt{periodIndFrom}: the source period indicator.
- \texttt{periodIndTo}: the target period indicator.

Examples of valid syntaxes

- \texttt{sum ( DS group all time\_agg ( Me, "A") )}
- \texttt{time\_agg ( "A", cast ( "2012Q1", time\_period , "YYYY\_Qq") )}
- \texttt{time\_agg("M", cast ( "2012-12-23", date, "YYYY-MM-DD") )}
- \texttt{time\_agg("M", DS1)}
- \texttt{ds\_2 := ds1[calc Me1 := time\_agg("M",Me1)]}

Seminantics for scalar operations

The operator \texttt{conv}erts a \texttt{time}, \texttt{date} or \texttt{time\_period} value from a smaller to a larger duration.

- \texttt{op}: \texttt{dataset} \{ identifier < \texttt{time} >_, identifier _* \}
  - \texttt{component<time>}
  - \texttt{time}

- \texttt{periodIndFrom}: \texttt{duration}
- \texttt{periodIndTo}: \texttt{duration}

Result type

- \texttt{op}: \texttt{dataset} \{ identifier < \texttt{time} >_, identifier _* \}
  - \texttt{component<time>}
  - \texttt{time}

Additional constraints

If \texttt{op} is a Data Set then it has exactly an Identifier of type \texttt{time}, \texttt{date} or \texttt{time\_period} and may have other Identifiers. It is only possible to convert smaller duration values to larger duration values (e.g. it is possible to convert \texttt{monthly} data to \texttt{annual} data but the contrary is not allowed).

Behaviour

The scalar version of this operator takes as input a \texttt{time}, \texttt{date} or \texttt{time\_period} value, converts it to \texttt{periodIndTo} and returns a scalar of the corresponding type. The Data Set version acts on a single Measure Data Set of type \texttt{time}, \texttt{date} or \texttt{time\_period} and returns a Data Set having the same structure. Finally, VTL also provides a component version, for use in combination with an aggregation operator, because the change of frequency requires an aggregation. In this case, the operator converts the \texttt{period\_indicator} of the data points (e.g., convert \texttt{monthly} data to \texttt{annual} data).

On \texttt{time} type, the operator maps the input value into the comprising larger regular interval, whose duration is the one specified by the \texttt{periodIndTo} parameter. On \texttt{date} type, the operator maps the input value into the comprising larger period, whose duration is the one specified by the \texttt{periodIndTo} parameter, which is conventionally represented either by the start or by the end date, according to the \texttt{first|last} parameter. On \texttt{time\_period} type, the operator maps the input value into the comprising larger time period specified by the \texttt{periodIndTo} parameter (the original period indicator is converted in the target one and the number of periods is adjusted correspondingly).

The input duration \texttt{periodIndFrom} is optional. In case of \texttt{time\_period} Data Points, the input duration can be inferred from the internal representation of the value. In case of \texttt{time} or \texttt{date} types, it is inferred by the implementation. Filters on input time series can be obtained with the \texttt{filter} clause.
Examples

Given the Data Set DS_1

<table>
<thead>
<tr>
<th>Id</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010Q1</td>
<td>A</td>
<td>20</td>
</tr>
<tr>
<td>2010Q2</td>
<td>A</td>
<td>20</td>
</tr>
<tr>
<td>2010Q3</td>
<td>A</td>
<td>20</td>
</tr>
<tr>
<td>2010Q1</td>
<td>B</td>
<td>50</td>
</tr>
<tr>
<td>2010Q2</td>
<td>B</td>
<td>50</td>
</tr>
<tr>
<td>2010Q1</td>
<td>C</td>
<td>10</td>
</tr>
<tr>
<td>2010Q2</td>
<td>C</td>
<td>10</td>
</tr>
</tbody>
</table>

Example 1:

$$DS_r := \text{sum}(DS_1) \text{ group all time_agg(“A”, _, Me_1)}$$

results in:

<table>
<thead>
<tr>
<th>Id</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>A</td>
<td>60</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>100</td>
</tr>
<tr>
<td>2010</td>
<td>C</td>
<td>20</td>
</tr>
</tbody>
</table>

Example 2:

$$DS_r := \text{time_agg(“Q”, cast(“2012M01”, time_period, “YYYY\MMM”))}$$

Returns: “2012Q1”.

Example 3:

The following example maps a date to quarter level, 2012 (end of the period).

$$\text{time_agg(“Q”, cast(“20120213”, date, “YYYYMMDD”), _, last)}$$

and produces a date value corresponding to the string “20120331”

Example 4:

The following example maps a date to year level, 2012 (beginning of the period).

$$\text{time_agg(cast(“A”, “2012M1”, date, “YYYYMMDD”), _, first)}$$

and produces a date value corresponding to the string “20120101”.

Actual time: \text{current_date}

Syntax

\text{current_date( )}

Input parameters

None

Examples of valid syntax
The operator `current_date` returns the current time as a `date` type.

This operator has no input parameters.

The operator return the current date

Examples

`cast ( current_date, string, "YYYY.MM.DD" )`
Union: union

Syntax
union ( dsList )
dsList ::= ds { , ds }*

Input parameters
dsList the list of Data Sets in the union

Examples of valid syntaxes
union ( ds2, ds3 )

Semantics for scalar operations
This operator does not perform scalar operations.

Input parameters type
ds :: dataset

Result type
result :: dataset

Additional constraints
All the Data Sets in dsList have the same Identifier, Measure and Attribute Components.

Behaviour
The union operator implements the union of functions (i.e., Data Sets). The resulting Data Set has the same Identifier, Measure and Attribute Components of the operand Data Sets specified in the dsList, and contains the Data Points belonging to any of the operand Data Sets.

The operand Data Sets can contain Data Points having the same values of the Identifiers. To avoid duplications of Data Points in the resulting Data Set, those Data Points are filtered by choosing the Data Point belonging to the left most operand Data Set. For instance, let’s assume that in union ( ds1, ds2 ) the operand ds1 contains a Data Point dp1 and the operand ds2 contains a Data Point dp2 such that dp1 has the same Identifiers values of dp2, then the resulting Data Set contains dp1 only.

The operator has the typical behaviour of the “Behaviour of the Set operators” (see the section “Typical behaviours of the ML Operators”).

The automatic Attribute propagation is not applied.

Examples
Given the operand Data Sets DS_1 and DS_2:

<table>
<thead>
<tr>
<th>DS_1</th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
<td>Id_2</td>
<td>Id_3</td>
<td>Id_4</td>
<td>Me_1</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>3</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>DS_2</th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
<td>Id_2</td>
<td>Id_3</td>
<td>Id_4</td>
<td>Me_1</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
### Example 1:

\[
DS_r := \text{union}(DS_1, DS_2)
\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>5</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>3</td>
</tr>
<tr>
<td>2012</td>
<td>N</td>
<td>Total</td>
<td>Total</td>
<td>23</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>5</td>
</tr>
</tbody>
</table>

Given the operand Data Sets DS_1 and DS_2:

#### DS_1

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>5</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>3</td>
</tr>
</tbody>
</table>

#### DS_2

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>23</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>5</td>
</tr>
</tbody>
</table>

### Example 2:

\[
DS_r := \text{union}(DS_1, DS_2)
\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>5</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>3</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>5</td>
</tr>
</tbody>
</table>

**Intersection:**

**Syntax**

\[
\text{intersect ( } ds\text{List } \text{) }
\]

\[
ds\text{List} ::= ds \{ , ds \}^*
\]

**Input Parameters**

- `dsList`: the list of Data Sets in the intersection
Examples of valid syntaxes
intersect ( ds2, ds3 )

Semantics for scalar operations
This operator cannot be applied to scalar values.

Input parameters type
ds :: dataset

Return type
result :: dataset

Additional constraints
All the Data Sets in dsList have the same Identifier, Measure and Attribute Components.

Behaviour
The intersect operator implements the intersection of functions (i.e., Data Sets). The resulting Data Set has the same Identifier, Measure and Attribute Components of the operand Data Sets specified in the dsList, and contains the Data Points belonging to all the operand Data Sets.

The operand Data Sets can contain Data Points having the same values of the Identifiers. To avoid duplications of Data Points in the resulting Data Set, those Data Points are filtered by choosing the Data Point belonging to the left most operand Data Set. For instance, let’s assume that in intersect ( ds1, ds2 ) the operand ds1 contains a Data Point dp1 and the operand ds2 contains a Data Point dp2 such that dp1 has the same Identifiers values of dp2, then the resulting Data Set contains dp1 only.

The operator has the typical behaviour of the “Behaviour of the Set operators” (see the section “Typical behaviours of the ML Operators”).

The automatic Attribute propagation is not applied.

Examples
Given the operand Data Sets DS_1 and DS_2:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>1</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>3</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>10</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>2</td>
</tr>
<tr>
<td>2011</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>40</td>
</tr>
</tbody>
</table>

Example 1: DS_r := intersect(DS_1,DS_2) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>2</td>
</tr>
</tbody>
</table>

Set difference: setdiff
**Syntax**

```plaintext
setdiff (ds1, ds2)
```

**Input parameters**

- `ds1` the first Data Set in the difference (the minuend)
- `ds2` the second Data Set in the difference (the subtrahend)

**Examples of valid syntaxes**

```plaintext
setdiff (ds2, ds3)
```

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type**

```plaintext
ds1, ds2 :: dataset
```

**Result type**

```plaintext
result :: dataset
```

**Additional constraints**

The operand Data Sets have the same Identifier, Measure and Attribute Components.

**Behaviour**

The operator implements the set difference of functions (i.e. Data Sets), interpreting the Data Points of the input Data Sets as the elements belonging to the operand sets, the minuend and the subtrahend, respectively. The operator returns one single Data Set, with the same Identifier, Measure and Attribute Components as the operand Data Sets, containing the Data Points that appear in the first Data Set but not in the second. In other words, for `setdiff (ds1, ds2)`, the resulting Dataset contains all the data points Data Point `dp1` of the operand `ds1` such that there is no Data Point `dp2` of `ds2` having the same values for homonym Identifier Components.

The operator has the typical behaviour of the “Behaviour of the Set operators” (see the section “Typical behaviours of the ML Operators”).

The automatic Attribute propagation is not applied.

**Examples**

Given the operand Data Sets `DS_1` and `DS_2`:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>10</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>20</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>30</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>40</td>
</tr>
<tr>
<td>2012</td>
<td>I</td>
<td>Total</td>
<td>Total</td>
<td>50</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>60</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>10</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>20</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>30</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>40</td>
</tr>
<tr>
<td>Year</td>
<td>ID</td>
<td>Total</td>
<td>Total</td>
<td>Value</td>
</tr>
<tr>
<td>------</td>
<td>----</td>
<td>-------</td>
<td>-------</td>
<td>-------</td>
</tr>
<tr>
<td>2012</td>
<td>I</td>
<td></td>
<td></td>
<td>50</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td></td>
<td></td>
<td>60</td>
</tr>
</tbody>
</table>

Example 1: \( DS_r := \text{setdiff}(DS_1, DS_2) \) results in:

<table>
<thead>
<tr>
<th>Year</th>
<th>ID</th>
<th>Total</th>
<th>Total</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>10</td>
</tr>
</tbody>
</table>

Given the operand Data Sets \( DS_1 \) and \( DS_2 \):

\[ DS_1 \]

<table>
<thead>
<tr>
<th>ID</th>
<th>Total</th>
<th>Total</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>R</td>
<td>M</td>
<td>2011</td>
<td>7</td>
</tr>
<tr>
<td>R</td>
<td>F</td>
<td>2011</td>
<td>10</td>
</tr>
<tr>
<td>R</td>
<td>T</td>
<td>2011</td>
<td>12</td>
</tr>
</tbody>
</table>

\[ DS_2 \]

<table>
<thead>
<tr>
<th>ID</th>
<th>Total</th>
<th>Total</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>R</td>
<td>M</td>
<td>2011</td>
<td>7</td>
</tr>
<tr>
<td>R</td>
<td>F</td>
<td>2011</td>
<td>10</td>
</tr>
</tbody>
</table>

Example 2: \( DS_r := \text{setdiff}(DS_1, DS_2) \) results in:

<table>
<thead>
<tr>
<th>ID</th>
<th>Total</th>
<th>Total</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>R</td>
<td>T</td>
<td>2011</td>
<td>12</td>
</tr>
</tbody>
</table>

Simmetric difference: \text{symdiff}

Syntax
\[ \text{symdiff}(ds1, ds2) \]

Input parameters
- \( ds1 \) the first Data Set in the difference
- \( ds2 \) the second Data Set in the difference

Examples of valid syntaxes
\text{symdiff}(ds_2, ds_3)

Semantics for scalar operations
This operator cannot be applied to scalar values.

Input parameters type
- \( ds1, ds2 :: \) dataset

Result type
result:: dataset

Additional constraints
The operand Data Sets have the same Identifier, Measure and Attribute Components.

Behaviour
The operator implements the symmetric set difference between functions (i.e. Data Sets), interpreting the Data Points of the input Data Sets as the elements in the operand Sets. The operator returns one Data Set, with the same Identifier, Measure and Attribute Components as the operand Data Sets, containing the Data Points that appear in the first Data Set but not in the second and the Data Points that appear in the second Data Set but not in the first one.

Data Points are compared to one another by Identifier Components. For symdiff (ds1, ds2), the resulting Data Set contains all the Data Points dp1 contained in ds1 for which there is no Data Point dp2 in ds2 with the same values for homonym Identifier components and all the Data Points dp2 contained in ds2 for which there is no Data Point dp1 in ds1 with the same values for homonym Identifier Components.

The operator has the typical behaviour of the "Behaviour of the Set operators" (see the section "Typical behaviours of the ML Operators"). The automatic Attribute propagation is not applied.

Examples
Given the operand Data Set DS_1 and DS_2:

<table>
<thead>
<tr>
<th>DS_1</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Id_1</strong></td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>DS_2</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Id_1</strong></td>
</tr>
<tr>
<td>2011</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
</tbody>
</table>

Example 1: DS_r := symdiff ( DS_1, DS_2 ) results in:

<table>
<thead>
<tr>
<th>DS_r</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Id_1</strong></td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2011</td>
</tr>
</tbody>
</table>
Hierarchical roll-up: hierarchy

**Syntax**

\[
\text{hierarchy} \ ( \ \text{op} \ , \ \text{hr} \ \{ \ \text{condition} \ \text{condComp} \ , \ \text{condComp} \} \ ^* \} \ \{ \ \text{rule} \ \text{ruleComp} \} \ \{ \ \text{mode} \} \ \{ \ \text{input} \} \ \{ \ \text{output} \} )
\]

- **mode**: non_null | non_zero | partial_null | partial_zero | always_null | always_zero
- **input**: dataset | rule | rule_priority
- **output**: computed | all

**Input parameters**

- **op**: the operand Data Set.
- **hr**: the hierarchical Ruleset to be applied.
- **condComp**: condComp is a Component of op to be associated (in positional order) to the conditioning Value Domains or Variables defined in hr (if any).
- **ruleComp**: ruleComp is the Identifier of op to be associated to the rule Value Domain or Variable defined in hr.
- **mode**: this parameter specifies how to treat the possible missing Data Points corresponding to the Code Items in the right side of a rule and which Data Points are produced in output. The meaning of the possible values of the parameter is explained below.
- **input**: this parameter specifies the source of the values used as input of the hierarchical rules. The meaning of the possible values of the parameter is explained below.
- **output**: this parameter specifies the content of the resulting Data Set. The meaning of the possible values of the parameter is explained below.

**Examples of valid syntaxes**

- `hierarchy ( DS1, HR1 rule Id_1 non_null all )`
- `hierarchy ( DS2, HR2 condition Comp_1, Comp_2 rule Id_3 non_zero rule computed )`

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type**

- **op**: dataset { measure<number> }_
- **hr**: name < hierarchical >
- **condComp**: name < component >
- **ruleComp**: name < dentifier >

**Result type**

- **result**: dataset { measure<number> }_

**Additional constraints**

If hr is defined on Value Domains then it is mandatory to specify the condition (if any) and the rule parameters. Moreover, the Components specified as condComp and ruleComp must belong to the operand op and must take values on the Value Domains corresponding, in positional order, to the ones specified in the condition and rule parameter of hr.

If hr is defined on Variables, the specification of condComp and ruleComp is not needed, but they can be specified all the same if it is desired to show explicitly in the invocation which are the involved Components: in this case, the condComp and ruleComp must be the same and in the same order as the Variables specified in in the condition and rule signatures of hr.

**Behaviour**

The hierarchy operator applies the rules of hr to op as specified in the parameters. The operator returns a Data Set with the same Identifiers and the same Measure as op. The Attribute propagation rule is applied on the groups of Data Points which contribute to the same Data Points of the result.

The behaviours relevant to the different options of the input parameters are the following.
First, the parameter input is considered to determine the source of the Data Points used as input of the dataset. For each Rule of the Ruleset and for each item on the right hand side of the Rule, the operator takes the input Data Points exclusively from the operand op.

For each Rule of the Ruleset and for each item on the right-hand side of the Rule:

- if the item is not defined as the result (left-hand side) of another Rule, the current Rule takes the input Data Points from the operand op.
- if the item is defined as the result of another Rule, the current Rule takes the input Data Points from the computed output of such other Rule; Points from the computed output of such other Rule;

For each Rule of the Ruleset and for each item on the right-hand side of the Rule:

- if the item is not defined as the result (left-hand side) of another rule, the current Rule takes the input Data Points from the operand op.
- if the item is defined as the result of another Rule, then:
  - if an expected input Data Point exists in the computed output of such other Rule and its Measure is not NULL, then the current Rule takes such Data Point;
  - if an expected input Data Point does not exist in the computed output of such other Rule or its measure is NULL, then the current Rule takes the Data Point from op (if any) having the same values of the Identifiers;

if the parameter input is not specified then it is assumed to be rule.

Then the parameter mode is considered, to determine the behaviour for missing Data Points and for the Data Points to be produced in the output. The possible options of the parameter mode and the corresponding behaviours are the following:

- non_null: the result Data Point is produced when its computed Measure value is not NULL (i.e., when no Data Point corresponding to the Code Items of the right side of the rule is missing or has NULL Measure value); in the calculation, the possible missing Data Points corresponding to the Code Items of the right side of the rule are considered existing and having a Measure value equal to NULL;
- non_zero: the result Data Point is produced when its computed Measure value is not equal to 0 (zero); the possible missing Data Points corresponding to the Code Items of the right side of the rule are considered existing and having a Measure value equal to 0;
- partial_null: the result Data Point is produced if at least one Data Point corresponding to the Code Items of the right side of the rule is found (whichever is its Measure value); the possible missing Data Points corresponding to the Code Items of the right side of the rule are considered existing and having a NULL Measure value;
- partial_zero: the result Data Point is produced if at least one Data Point corresponding to the Code Items of the right side of the rule is found (whichever is its Measure value); the possible missing Data Points corresponding to the Code Items of the right side of the rule are considered existing and having a Measure value equal to 0 (zero);
- always_null: the result Data Point is produced in any case; the possible missing Data Points corresponding to the Code Items of the right side of the rule are considered existing and having a Measure value equal to NULL;
- always_zero: the result Data Point is produced in any case; the possible missing Data Points corresponding to the Code Items of the right side of the rule are considered existing and having a Measure value equal to 0 (zero);

If the parameter mode is not specified, then it is assumed to be non_null

The following table summarizes the behaviour of the options of the parameter "mode".

<table>
<thead>
<tr>
<th>OPTION of the MODE PARAMETER:</th>
<th>Missing Data Points are considered:</th>
<th>Null Data Points are considered:</th>
<th>Condition for evaluating the rule</th>
<th>Returned Data Points</th>
</tr>
</thead>
<tbody>
<tr>
<td>Non_null</td>
<td>NULL</td>
<td>NULL</td>
<td>If all the involved Data Points are not NULL</td>
<td>Only not NULL Data Points (Zeros are returned too)</td>
</tr>
<tr>
<td>Non_zero</td>
<td>Zero</td>
<td>NULL</td>
<td>If at least one of the involved Data Points is &lt;&gt; zero</td>
<td>Only not zero Data Points (NULLS are returned too)</td>
</tr>
</tbody>
</table>
If at least one of the involved Data Points is not NULL

Data Points of any value (NULL, not NULL and zero too)

If at least one of the involved Data Points is not NULL

Data Points of any value (NULL, not NULL and zero too)

Always

Data Points of any value (NULL, not NULL and zero too)

Finally the parameter output is considered, to determine the content of the resulting Data Set. The possible options of the parameter output and the corresponding behaviours are the following:

- computed: the resulting Data Set contains only the set of Data Points computed according to the Ruleset
- all: the resulting Data Set contains the union between the set of Data Points "R" computed according to the Ruleset and the set of Data Points of op that have different combinations of values for the Identifiers. In other words, the result is the outcome of the following (virtual) expression:
  \[
  \text{union ( setdiff (op , R) , R )}
  \]

If the parameter output is not specified then it is assumed to be computed.

### Examples

Given the following hierarchical ruleset:

```plaintext
define hierarchical ruleset HR_1 ( valuedomain rule VD_1 ) is
  A = J + K + L
  ; B = M + N + O
  ; C = P + Q
  ; D = R + S
  ; E = T + U + V
  ; F = Y + W + Z
  ; G = B + C
  ; H = D + E
  ; I = D + G
end hierarchical ruleset
```

And given the operand Data Set DS_1 (where At_1 is viral and the propagation rule says that the alphabetic order prevails the NULL prevails on the alphabetic characters and the Attribute value for missing Data Points is assumed as NULL):

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>M</td>
<td>2</td>
<td>Dx</td>
</tr>
<tr>
<td>2010</td>
<td>N</td>
<td>5</td>
<td>Pz</td>
</tr>
<tr>
<td>2010</td>
<td>O</td>
<td>4</td>
<td>Pz</td>
</tr>
<tr>
<td>2010</td>
<td>P</td>
<td>7</td>
<td>Pz</td>
</tr>
<tr>
<td>2010</td>
<td>Q</td>
<td>-7</td>
<td>Pz</td>
</tr>
<tr>
<td>2010</td>
<td>S</td>
<td>3</td>
<td>Ay</td>
</tr>
<tr>
<td>2010</td>
<td>T</td>
<td>9</td>
<td>Bq</td>
</tr>
<tr>
<td>2010</td>
<td>U</td>
<td>NULL</td>
<td>Nj</td>
</tr>
</tbody>
</table>
Example 1:  \[ DS_r := \text{hierarchy} \left( DS_1, HR_1 \rule{Id_2}{\text{non\_null}} \right) \]

\[
\begin{array}{|c|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 & \text{At}_1 \\
\hline
2010 & B & 11 & \text{Dx} \\
2010 & C & 0 & \text{Pz} \\
2010 & G & 19 & \text{Dx} \\
\hline
\end{array}
\]

Example 2:  \[ DS_r := \text{hierarchy} \left( DS_1, HR_1 \rule{Id_2}{\text{non\_zero}} \right) \]

\[
\begin{array}{|c|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 & \text{At}_1 \\
\hline
2010 & B & 11 & \text{Dx} \\
2010 & D & 3 & \text{NULL} \\
2010 & E & \text{NULL} & \text{Bq} \\
2010 & G & 11 & \text{Dx} \\
2010 & H & \text{NULL} & \text{NULL} \\
2010 & I & 14 & \text{NULL} \\
\hline
\end{array}
\]

Example 2:  \[ DS_r := \text{hierarchy} \left( DS_1, HR_1 \rule{Id_2}{\text{partial\_null}} \right) \]

\[
\begin{array}{|c|c|c|c|}
\hline
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 & \text{At}_1 \\
\hline
2010 & B & 11 & \text{Dx} \\
2010 & C & 0 & \text{Pz} \\
2010 & D & \text{NULL} & \text{NULL} \\
2010 & E & \text{NULL} & \text{Bq} \\
2010 & G & 11 & \text{Dx} \\
2010 & H & \text{NULL} & \text{NULL} \\
2010 & I & \text{NULL} & \text{NULL} \\
\hline
\end{array}
\]
The following table lists the operators that can be invoked in the Aggregate or in the Analytic invocations described below and their main characteristics.

<table>
<thead>
<tr>
<th>Operator</th>
<th>Description</th>
<th>Allowed</th>
<th>Type of the resulting Measure</th>
<th>Type of the operand Measures</th>
</tr>
</thead>
<tbody>
<tr>
<td>count</td>
<td>number of Data Points</td>
<td>Aggregate, Analytic</td>
<td>integer</td>
<td>any</td>
</tr>
<tr>
<td>min</td>
<td>minimum value of a set of values</td>
<td>Aggregate, Analytic</td>
<td>any</td>
<td>any</td>
</tr>
<tr>
<td>max</td>
<td>maximum value of a set of values</td>
<td>Aggregate, Analytic</td>
<td>any</td>
<td>any</td>
</tr>
<tr>
<td>median</td>
<td>median value of a set of numbers</td>
<td>Aggregate, Analytic</td>
<td>number</td>
<td>number</td>
</tr>
<tr>
<td>sum</td>
<td>sum of a set of numbers</td>
<td>Aggregate, Analytic</td>
<td>number</td>
<td>number</td>
</tr>
<tr>
<td>avg</td>
<td>average value of a set of numbers</td>
<td>Aggregate, Analytic</td>
<td>number</td>
<td>number</td>
</tr>
<tr>
<td>stddev_pop</td>
<td>population standard deviation of a set of numbers</td>
<td>Aggregate, Analytic</td>
<td>number</td>
<td>number</td>
</tr>
<tr>
<td>stddev_samp</td>
<td>sample standard deviation of a set of numbers</td>
<td>Aggregate, Analytic</td>
<td>number</td>
<td>number</td>
</tr>
<tr>
<td>var_pop</td>
<td>population variance of a set of numbers</td>
<td>Aggregate, Analytic</td>
<td>number</td>
<td>number</td>
</tr>
<tr>
<td>var_samp</td>
<td>sample variance of a set of numbers</td>
<td>Aggregate, Analytic</td>
<td>number</td>
<td>number</td>
</tr>
<tr>
<td>first_value</td>
<td>first value in an ordered set of values</td>
<td>Analytic</td>
<td>any</td>
<td>any</td>
</tr>
<tr>
<td>last_value</td>
<td>last value in an ordered set of values</td>
<td>Analytic</td>
<td>any</td>
<td>any</td>
</tr>
<tr>
<td>lag</td>
<td>in an ordered set of Data Points, it returns the value(s)</td>
<td>Analytic</td>
<td>any</td>
<td>any</td>
</tr>
<tr>
<td>lead</td>
<td>in an ordered set of Data Points, it returns the value(s)</td>
<td>Analytic</td>
<td>any</td>
<td>any</td>
</tr>
<tr>
<td>rank</td>
<td>rank (order number) of a Data Point in an ordered set of</td>
<td>Analytic</td>
<td>integer</td>
<td>any</td>
</tr>
</tbody>
</table>
Aggregate invocation

Syntax

in a Data Set expression:

\[ \text{aggregateOperator ( firstOperand \{ , additionalOperand \}*) \{ groupingClause \} } \]

in a Component expression within an aggr clause:

\[ \text{aggregateOperator ( firstOperand \{ , additionalOperand \}*) \{ groupingClause \} } \]

\[ \text{aggregateOperator ::= avg | count | max | median | min | stddev_pop} \]

\[ | stddev_samp | sum | var_pop | var_samp \]

\[ \text{groupingClause ::= \{ group by groupingId \{, groupingId \}* \}} \]

\[ | \text{ group except groupingId \{, groupingId \}*} \]

\[ | \text{ group all conversionExpr} \}

\[ | \text{ having havingCondition} \}

Input Parameters

- **aggregateOperator**: the keyword of the aggregate operator to invoke (e.g., `avg`, `count`, `max` ...)
- **firstOperand**: the first operand of the invoked aggregate operator (a Data Set for an invocation at Data Set level or a Component of the input Data Set for an invocation at Component level within a `aggr` operator or a `aggr` clause in a join operation)
- **additionalOperand**: an additional operand (if any) of the invoked operator. The various operators can have a different number of parameters. The number of parameters, their types and if they are mandatory or optional depend on the invoked operator
- **groupingClause**: the following alternative grouping options:
  - **group by**: the Data Points are grouped by the values of the specified Identifiers (`groupingId`). The Identifiers not specified are dropped in the result.
  - **group except**: the Data Points are grouped by the values of the Identifiers not specified as `groupingId`. The Identifiers specified as `groupingId` are dropped in the result.
  - **group all**: converts the values of an Identifier Component using `conversionExpr` and keeps all the resulting Identifiers.
- **groupingId**: Identifier Component to be kept (in the `group by` clause) or dropped (in the `group except` clause).
- **conversionExpr**: specifies a conversion operator (e.g., `time_agg`) to convert data from finer to coarser granularity. The conversion operator is applied on an Identifier of the operand Data Set op.
- **havingCondition**: a condition (boolean expression) at component level, having only Components of the input Data Sets as operands (and possibly constants), to be fulfilled by the groups of Data Points: only groups for which `havingCondition` evaluates to `TRUE` appear in the result. The `havingCondition` refers to the groups specified through the `groupingClause`, therefore it must invoke aggregate operators (e.g., `avg`, `count`, `max` ..., see also the corresponding sections). A correct example of `havingCondition` is:

\[ \text{max(obs_value) < 1000} \]

while the condition `obs_value < 1000` is not a right `havingCondition`, because it refers to the values of single Data Points and not to the groups. The count operator is used in a `havingCondition` without parameters, e.g.:  

\[ \text{sum ( ds group by id1 having count ( ) >= 10 )} \]

Examples of valid syntaxes

- `avg ( DS_1 )`
- `avg ( DS_1 group by Id_1, Id_2 )`
The aggregate operators cannot be applied to scalar values.

**Input parameters type**

- `firstOperand ::` dataset | component
- `additionalOperand ::` see the type of the additional parameter (if any) of the invoked `aggregateOperator`. The aggregate operators and their parameters are described in the following sections.
- `groupingIId ::` name < identifier >
- `conversionExpr ::` identifier
- `havingCondition ::` component<boolean>

**Result type:**

- `result ::` dataset | component

**Additional constraints**

The Aggregate invocation cannot be nested in other Aggregate or Analytic invocations.

The aggregate operations at component level can be invoked within the `aggr` clause, both as part of a join operator and the `aggr` operator (see the parameter `aggrExpr` of those operators).

The basic scalar types of `firstOperand` and `additionalOperand` (if any) must be compliant with the specific basic scalar types required by the invoked operator (the required basic scalar types are described in the table at the beginning of this chapter and in the sections of the various operators below).

The `conversionExpr` parameter applies just one conversion operator to just one Identifier belonging to the input Data Set. The basic scalar type of the Identifier must be compatible with the basic scalar type of the conversion operator.

If the grouping clause is omitted, then all the input Data Points are aggregated in a single group and the clause returns a Data Set that contains a single Data Point and has no Identifiers.

**Behaviour**

The `aggregateOperator` is applied as usual to all the measures of the `firstOperand` Data Set (if invoked at Data Set level) or to the `firstOperand` Component of the input Data Set (if invoked at Component level). In both cases, the operator calculates the required aggregated values for groups of Data Points of the input Data Set. The groups of Data Points to be aggregated are specified through the `groupingClause`, which allows the following alternative options.

- `group by` the Data Points are grouped by the values of the specified Identifiers. The Identifiers not specified are dropped in the result.
- `group except` the Data Points are grouped by the values of the Identifiers not specified in the clause. The specified Identifiers are dropped in the result.
- `group all` converts an Identifier Component using `conversionExpr` and keeps all the Identifiers.

The `having` clause is used to filter groups in the result by means of an aggregate condition evaluated on the single groups (for example the minimum number of rows in the group).

If no grouping clause is specified, then all the input Data Points are aggregated in a single group and the operator returns a Data Set that contains a single Data Point and has no Identifiers.

For the invocation at Data Set level, the resulting Data Set has the same Measures as the operand. For the invocation at Component level, the resulting Data Set has the Measures explicitly calculated (all the other Measures are dropped because no aggregation behaviour is specified for them).

For invocation at Data Set level, the Attribute propagation rule is applied. For invocation at Component level, the Attributes calculated within the `aggr` clause are maintained in the result; for all the other Attributes that are defined as `viral`, the Attribute propagation rule is applied (for the semantics, see the Attribute Propagation Rule section in the User Manual).

As mentioned, the Aggregate invocation at component level can be done within the `aggr` clause, both as part of a Join operator and the `aggr` operator (see the parameter `aggrExpr` of those operators), therefore, for a better comprehension of the behaviour at Component level, see also those operators.
Examples

Given the Data Set DS_1

<table>
<thead>
<tr>
<th>id_1</th>
<th>id_2</th>
<th>id_3</th>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>E</td>
<td>XX</td>
<td>20</td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>B</td>
<td>XX</td>
<td>1</td>
<td>H</td>
</tr>
<tr>
<td>2010</td>
<td>R</td>
<td>XX</td>
<td>1</td>
<td>A</td>
</tr>
<tr>
<td>2010</td>
<td>F</td>
<td>YY</td>
<td>23</td>
<td></td>
</tr>
<tr>
<td>2011</td>
<td>E</td>
<td>XX</td>
<td>20</td>
<td>P</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>ZZ</td>
<td>1</td>
<td>N</td>
</tr>
<tr>
<td>2011</td>
<td>R</td>
<td>YY</td>
<td>-1</td>
<td>P</td>
</tr>
<tr>
<td>2011</td>
<td>F</td>
<td>XX</td>
<td>20</td>
<td>Z</td>
</tr>
<tr>
<td>2012</td>
<td>L</td>
<td>ZZ</td>
<td>40</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>E</td>
<td>YY</td>
<td>30</td>
<td>P</td>
</tr>
</tbody>
</table>

**Example 1:**  
\[ DS_{r} := \text{avg}\ (\ DS_{1}\ \text{group by}\ id_{1}) \]  
provided that \(At_{1}\) is non viral, results in:

<table>
<thead>
<tr>
<th>id_1</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>11.25</td>
</tr>
<tr>
<td>2011</td>
<td>10</td>
</tr>
<tr>
<td>2012</td>
<td>35</td>
</tr>
</tbody>
</table>

Note: the example above can be rewritten equivalently in the following forms:

\[ DS_{r} := \text{avg}\ (\ DS_{1}\ \text{group except}\ id_{2},\ id_{3}) \]
\[ DS_{r} := \text{avg}\ (\ DS_{1}\#Me_{1}\ \text{group by}\ id_{1}) \]

**Example 2:**  
\[ DS_{r} := \text{sum}\ (\ DS_{1}\ \text{group by}\ id_{1},\ id_{3}) \]  
provided that \(At_{1}\) is non viral, results in:

<table>
<thead>
<tr>
<th>id_1</th>
<th>id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>XX</td>
<td>22</td>
</tr>
<tr>
<td>2010</td>
<td>YY</td>
<td>23</td>
</tr>
<tr>
<td>2011</td>
<td>XX</td>
<td>40</td>
</tr>
<tr>
<td>2011</td>
<td>ZZ</td>
<td>1</td>
</tr>
<tr>
<td>2011</td>
<td>YY</td>
<td>-1</td>
</tr>
<tr>
<td>2012</td>
<td>ZZ</td>
<td>40</td>
</tr>
<tr>
<td>2012</td>
<td>YY</td>
<td>30</td>
</tr>
</tbody>
</table>

**Example 3:**  
\[ DS_{r} := \text{avg}\ (\ DS_{1}) \]  
provided that \(At_{1}\) is non viral results in:
Example 4: \[ DS_r := DS_1 \ [ \text{aggr} \ Me_2 := \max ( Me_1 ) \ , \ Me_3 := \min ( Me_1 ) \ \text{group by} \ \text{Id}_1 \] \]

provided that \( \text{At}_1 \) is viral and the first letter in alphabetic order prevails and NULL prevails on all the other characters, results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Me_2</th>
<th>Me_3</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>23</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>2011</td>
<td>20</td>
<td>-1</td>
<td>N</td>
</tr>
<tr>
<td>2012</td>
<td>40</td>
<td>30</td>
<td>P</td>
</tr>
</tbody>
</table>

**Analytic invocation**

*Syntax*

\[
\text{analyticOperator} \ ( \text{firstOperand} \{ , \text{additionalOperand} \}^* \ \text{over} ( \text{analyticClause} ) )
\]

\[
\text{analyticOperator} ::= \text{avg} \mid \text{count} \mid \text{max} \mid \text{median} \mid \text{min} \mid \text{stddev}_\text{pop} \\
\mid \text{stddev}_\text{samp} \mid \text{sum} \mid \text{var}_\text{pop} \mid \text{var}_\text{samp} \\
\mid \text{first}_\text{value} \mid \text{lag} \mid \text{last}_\text{value} \mid \text{lead} \mid \text{rank} \mid \text{ratio}_\text{to}_\text{report}
\]

\[
\text{analyticClause} ::= \{ \text{partitionClause} \} \ ( \text{orderClause} ) \ ( \text{windowClause} )
\]

\[
\text{partitionClause} ::= \text{partition} \ \text{by} \ \text{identifier} \{ , \text{identifier} \}^*
\]

\[
\text{orderClause} ::= \text{order} \ \text{by} \ \text{component} \{ \ \text{asc} \mid \text{desc} \ \} \{ , \ \text{component} \{ \ \text{asc} \mid \text{desc} \ \} \}^*
\]

\[
\text{windowClause} ::= \{ \ \text{data points} \mid \text{range} \}^1 \ \text{between} \ \text{limitClause} \ \text{and} \ \text{limitClause}
\]

\[
\text{limitClause} ::= \{ \ \text{num} \ \text{preceding} \mid \text{num} \ \text{following} \mid \text{current data point} \mid \text{unbounded} \ \text{preceding} \mid \text{unbounded} \ \text{following} \}^1
\]

*Parameters*

\[
\text{analyticOperator} \ the \ keyword \ of \ the \ analytic \ operator \ to \ invoke \ (e.g., \text{avg} , \text{count} , \text{max} \ ...)
\]

\[
\text{firstOperand} \ the \ first \ operand \ of \ the \ invoked \ analytic \ operator \ (a \ Data \ Set \ for \ an \ invocation \ at \ Data \ Set \ level \ or \ a \ Component \ of \ the \ input \ Data \ Set \ for \ an \ invocation \ at \ Component \ level \ within \ a \ \text{calc} \ \text{operator} \ or \ a \ \text{calc} \ \text{clause} \ \text{in} \ \text{a} \ \text{join} \ \text{operation})
\]

\[
\text{additionalOperand} \ an \ additional \ operand \ (if \ any) \ of \ the \ invoked \ operator. \ The \ various \ operators \ can \ have \ a \ different \ number \ of \ parameters. \ The \ number \ of \ parameters, \ their \ types \ and \ if \ they \ are \ mandatory \ or \ optional \ depend \ on \ the \ invoked \ operator
\]

\[
\text{analyticClause} \ the \ keyword \ that \ specifies \ the \ analytic \ behaviour
\]

\[
\text{partitionClause} \ the \ keyword \ that \ specifies \ how \ to \ partition \ Data \ Points \ in \ groups \ to \ be \ analysed \ separately. \ The \ input \ Data \ Set \ is \ partitioned \ according \ to \ the \ values \ of \ one \ or \ more \ Identifier \ Components. \ If \ the \ clause \ is \ omitted, \ then \ the \ Data \ Set \ is \ partitioned \ by \ the \ Identifier \ Components \ that \ are \ not \ specified \ in \ the \ \text{orderClause}.
\]

\[
\text{orderClause} \ the \ keyword \ that \ specifies \ how \ to \ order \ the \ Data \ Points. \ The \ input \ Data \ Set \ is \ ordered \ according \ to \ the \ values \ of \ one \ or \ more \ Components, \ in \ ascending \ order \ if \ \text{asc} \ is \ specified, \ in \ descending \ order \ if \ \text{desc} \ is \ specified, \ by \ default \ in \ ascending \ order \ if \ the \ \text{asc} \ and \ \text{desc} \ keywords \ are \ omitted.
\]

\[
\text{windowClause} \ the \ keyword \ that \ specifies \ how \ to \ apply \ a \ sliding \ window \ on \ the \ ordered \ Data \ Points. \ The \ keyword \ \text{data points} \ means \ that \ the \ sliding \ window \ includes \ a \ certain \ number \ of \ Data \ Points \ before \ and \ after \ the \ current \ Data \ Point \ in \ the \ order \ given \ by \ the \ \text{orderClause}. \ The \ keyword \ \text{range} \ means \ that \ the \ sliding \ windows \ includes \ all \ the \ Data \ Points \ whose \ values \ are \ in \ a \ certain \ range \ in \ respect \ to \ the \ value, \ for \ the \ current \ Data \ Point, \ of \ the \ Measure \ which \ the \ analytic \ is \ applied \ to.
\]
limitClause clause that can specify either the lower or the upper boundaries of the sliding window. Each boundary is specified in relationship either to the whole partition or to the current data point under analysis by using the following keywords:

- **unbounded preceding** means that the sliding window starts at the first Data Point of the partition (it make sense only as the first limit of the window)
- **unbounded following** indicates that the sliding window ends at the last Data Point of the partition (it makes sense only as the second limit of the window)
- **current data point** specifies that the window starts or ends at the current Data Point.
- **num preceding** specifies either the number of data points to consider preceding the current data point in the order given by the orderClause (when data points is specified in the window clause), or the maximum difference to consider, as for the Measure which the analytic is applied to, between the value of the current Data Point and the generic other Data Point (when range is specified in the windows clause).
- **num following** specifies either the number of data points to consider following the current data point in the order given by the orderClause (when data points is specified in the window clause), or the maximum difference to consider, as for the Measure which the analytic is applied to, between the values of the generic other Data Point and the current Data Point (when range is specified in the windows clause).

If the whole windowClause is omitted then the default is data points between unbounded preceding and current data point.

**Examples of valid syntaxes**

```
sum ( DS_1 over ( partition by  Id_1  order by  Id_2  ) ) 
sum ( DS_1 over ( order by  Id_2  ) ) 
avg ( DS_1 over ( order by  Id_2  ) ) 
DS_1 [ calc M1 := sum ( Me_1 over ( order by  Id_1  ) ) ]
```

**Semantics for scalar operations**

The analytic operators cannot be applied to scalar values.

```
Input parameters type
```

```
firstOperand :: dataset | component
additionalOperand :: see the type of the additional parameter (if any) of the invoked operator. The operators and their parameters are described in the following sections.
identifier :: name < identifier >
component :: name < component >
um :: integer
```

```
Result type
```

```
result :: dataset | component
```

**Additional constraints**

The analytic invocation cannot be nested in other Aggregate or Analytic invocations.

The analytic operations at component level can be invoked within the calc clause, both as part of a Join operator and the calc operator (see the parameter calcExpr of those operators).

The basic scalar types of firstOperand and additionalOperand (if any) must be compliant with the specific basic scalar types required by the invoked operator (the required basic scalar types are described in the table at the beginning of this chapter and in the sections of the various operators below).
The analytic Operator is applied as usual to all the Measures of the input Data Set (if invoked at Data Set level) or to the specified Component of the input Data Set (if invoked at Component level). In both cases, the operator calculates the desired output values for each Data Point of the input Data Set.

The behaviour of the analytic operations can be procedurally described as follows:

- The Data Points of the input Data Set are first partitioned (according to partitionBy) and then ordered (according to orderBy).
- The operation is performed for each Data Point (named “current Data Point”) of the input Data Set. For each input Data Point, one output Data Point is returned, having the same values of the Identifiers. The analytic operator is applied to a “window” which includes a set of Data Points of the input Data Set and returns the values of the Measure(s) of the output Data Point.
- If windowClause is not specified, then the set of Data Points which contribute to the analytic operation is the whole partition which the current Data Point belongs to.
- If windowClause is specified, then the set of Data Points is the one specified by windowClause (see windowsClause and LimitClause explained above).

For the invocation at Data Set level, the resulting Data Set has the same Measures as the input Data Set firstOperand. For the invocation at Component level, the resulting Data Set has the Measures of the input Data Set plus the Measures explicitly calculated through the calc clause.

For the invocation at Data Set level, the Attribute propagation rule is applied. For invocation at Component level, the Attributes calculated within the calc clause are maintained in the result; for all the other Attributes that are defined as viral, the Attribute propagation rule is applied (for the semantics, see the Attribute Propagation Rule section in the User Manual).

As mentioned, the Analytic invocation at component level can be done within the calc clause, both as part of a Join operator and the calc operator (see the parameter aggrCalc of those operators), therefore, for a better comprehension fo the behaviour at Component level, see also those operators.

Examples

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>E</td>
<td>XX</td>
<td>5</td>
</tr>
<tr>
<td>2010</td>
<td>B</td>
<td>XX</td>
<td>-3</td>
</tr>
<tr>
<td>2010</td>
<td>R</td>
<td>XX</td>
<td>9</td>
</tr>
<tr>
<td>2010</td>
<td>E</td>
<td>YY</td>
<td>13</td>
</tr>
<tr>
<td>2011</td>
<td>E</td>
<td>XX</td>
<td>11</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>ZZ</td>
<td>7</td>
</tr>
<tr>
<td>2011</td>
<td>E</td>
<td>YY</td>
<td>-1</td>
</tr>
<tr>
<td>2011</td>
<td>F</td>
<td>XX</td>
<td>0</td>
</tr>
<tr>
<td>2012</td>
<td>L</td>
<td>ZZ</td>
<td>-2</td>
</tr>
<tr>
<td>2012</td>
<td>E</td>
<td>YY</td>
<td>3</td>
</tr>
</tbody>
</table>

Example1:

DS_r := sum ( DS_1 over ( order by Id_1, Id_2, Id_3 data points between 1 preceding and 1 following ) )

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>E</td>
<td>XX</td>
<td>5</td>
</tr>
<tr>
<td>2010</td>
<td>B</td>
<td>XX</td>
<td>-3</td>
</tr>
<tr>
<td>2010</td>
<td>R</td>
<td>XX</td>
<td>9</td>
</tr>
<tr>
<td>2010</td>
<td>E</td>
<td>YY</td>
<td>13</td>
</tr>
<tr>
<td>2011</td>
<td>E</td>
<td>XX</td>
<td>11</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>ZZ</td>
<td>7</td>
</tr>
<tr>
<td>2011</td>
<td>E</td>
<td>YY</td>
<td>-1</td>
</tr>
<tr>
<td>2011</td>
<td>F</td>
<td>XX</td>
<td>0</td>
</tr>
<tr>
<td>2012</td>
<td>L</td>
<td>ZZ</td>
<td>-2</td>
</tr>
<tr>
<td>2012</td>
<td>E</td>
<td>YY</td>
<td>3</td>
</tr>
</tbody>
</table>
### Counting the number of data points: \( \text{count} \)

#### Aggregate syntax

```
\text{count} ( \text{dataset} ( \text{groupingClause} ) ) \quad (\text{in a Data Set expression})
```

```
\text{count} ( \text{component} ( \text{groupingClause} ) ) \quad (\text{in a Component expression within an aggr clause})
```

```
\text{count} ( ) \quad (\text{in an having clause})
```

#### Analytic syntax

```
\text{count} ( \text{dataset over ( analyticClause ) } ) \quad (\text{in a Data Set expression})
```

```
\text{count} ( \text{component over ( analyticClause ) } ) \quad (\text{in a Component expression within a calc clause})
```

#### Input parameters

- `dataset`: the operand Data Set
- `component`: the operand Component
- `groupingClause`: see Aggregate invocation
- `analyticClause`: see Analytic invocation

#### Examples of valid syntaxes

See Aggregate and Analytic invocations above, at the beginning of the section.

#### Semantics for scalar operations

This operator cannot be applied to scalar values.

#### Input parameters type

- `dataset :: dataset`
- `component :: component`

#### Result type

```
\text{result :: dataset} \{ \text{measure<integer>} int\_var \}
```

```
\text{result :: component<integer>}
```

#### Additional constraints

None.

#### Behaviour

The operator returns the number of the input Data Points.

For other details, see Aggregate and Analytic invocations.

#### Examples

Given the Data Set `DS_1`:
Example 1:  
\[ DS_r := \text{count} \left( DS_1 \text{ group by } \text{Id}_1 \right) \]
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Int_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>3</td>
</tr>
<tr>
<td>2012</td>
<td>2</td>
</tr>
</tbody>
</table>

Example 1: use of count in a having clause:

\[ DS_r := \text{sum} \left( DS_1 \text{ group by } \text{Id}_1 \text{ having count()} > 2 \right) \]
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Int_var</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>3</td>
</tr>
</tbody>
</table>

Minimum value : \( \text{min} \)

**Aggregate syntax**

\[ \text{min} \left( \text{dataset} \right) \text{ groupingClause } \]
\[ \text{(in a Data Set expression)} \]

\[ \text{min} \left( \text{component} \right) \text{ groupingClause } \]
\[ \text{(in a Component expression within an aggr clause)} \]

**Analytic syntax**

\[ \text{min} \left( \text{dataset over} \right) \text{ analyticClause } \]
\[ \text{(in a Data Set expression)} \]

\[ \text{min} \left( \text{component over} \right) \text{ analyticClause } \]
\[ \text{(in a Component expression within a calc clause)} \]

**Input parameters**

- **dataset** the operand Data Set
- **component** the operand Component
- **groupingClause** see Aggregate invocation
- **analyticClause** see Analytic invocation

**Examples of valid syntaxes**

See Aggregate and Analytic invocations above, at the beginning of the section.

**Semantics for scalar operations**

This operator cannot be applied to scalar values.
**Input parameters type**

dataset :: dataset
component :: component

**Result type**

result :: dataset |
component

**Additional constraints**

None.

**Behaviour**

The operator returns the minimum value of the input values.
For other details, see Aggregate and Analytic invocations.

**Examples**

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>A</td>
<td>XX</td>
<td>3</td>
</tr>
<tr>
<td>2011</td>
<td>A</td>
<td>YY</td>
<td>5</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>YY</td>
<td>7</td>
</tr>
<tr>
<td>2012</td>
<td>A</td>
<td>XX</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>YY</td>
<td>4</td>
</tr>
</tbody>
</table>

Example 1: 

\[
\text{DS}_r := \min ( \text{DS}_1 \text{ group by Id}_1 )
\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>3</td>
</tr>
<tr>
<td>2012</td>
<td>2</td>
</tr>
</tbody>
</table>

**Maximum value:** \( \max \)

**Aggregate syntax**

\[
\max ( \text{dataset} \{ \text{groupingClause} \} ) \\
(\text{in a Data Set expression})
\]

\[
\max ( \text{component} \{ \text{groupingClause} \} ) \\
(\text{in a Component expression within an aggr clause})
\]

**Analytic syntax**

\[
\max ( \text{dataset} \over ( \text{analyticClause} ) ) \\
(\text{in a Data Set expression})
\]

\[
\max ( \text{component} \over ( \text{analyticClause} ) ) \\
(\text{in a Component expression within a calc clause})
\]

**Input parameters**

dataset the operand Data Set
component the operand Component
groupingClause see Aggregate invocation
analyticClause see Analytic invocation
Examples of valid syntaxes
See Aggregate and Analytic invocations above, at the beginning of the section.

Semantics for scalar operations
This operator cannot be applied to scalar values.

Input parameters type
dataset :: dataset
component :: component

Result type
result :: dataset
component

Additional constraints
None.

Behaviour
The operator returns the maximum of the input values.
For other details, see Aggregate and Analytic invocations.

Examples
Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>A</td>
<td>XX</td>
<td>3</td>
</tr>
<tr>
<td>2011</td>
<td>A</td>
<td>YY</td>
<td>5</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>YY</td>
<td>7</td>
</tr>
<tr>
<td>2012</td>
<td>A</td>
<td>XX</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>YY</td>
<td>4</td>
</tr>
</tbody>
</table>

Example 1: DS_r := max ( DS_1 group by Id_1 ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>7</td>
</tr>
<tr>
<td>2012</td>
<td>4</td>
</tr>
</tbody>
</table>

Median value : median

Aggregate syntax
median ( dataset { groupingClause } ) (in a Data Set expression)
median ( component ) { groupingClause } (in a Component expression within an aggr clause)

Analytic syntax
median ( dataset over ( partitionClause ) ) (in a Data Set expression)
median ( component over ( partitionClause ) ) (in a Component expression within a calc clause)
Input parameters

- **dataset**: the operand Data Set
- **component**: the operand Component
- **groupingClause**: see Aggregate invocation
- **analyticClause**: see Analytic invocation

Examples of valid syntaxes
See Aggregate and Analytic invocations above, at the beginning of the section.

Semantics for scalar operations
This operator cannot be applied to scalar values.

Input parameters type

- **dataset**: dataset {measure<number> _+}
- **component**: component<number>

Result type

- **result**: dataset { measure<number> _+ }
  - | component<number>

Additional constraints
None.

Behaviour
The operator returns the median value of the input values.
For other details, see Aggregate and Analytic invocations.

Examples

Given the Data Set **DS_1**:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>A</td>
<td>XX</td>
<td>3</td>
</tr>
<tr>
<td>2011</td>
<td>A</td>
<td>YY</td>
<td>5</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>YY</td>
<td>7</td>
</tr>
<tr>
<td>2012</td>
<td>A</td>
<td>XX</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>YY</td>
<td>4</td>
</tr>
</tbody>
</table>

**Example 1:** DS_r := median ( DS_1 group by Id_1 )
results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>5</td>
</tr>
<tr>
<td>2012</td>
<td>3</td>
</tr>
</tbody>
</table>

Sum:

Sum syntax

- **sum**: sum (dataset {groupingClause}) (in a Data Set expression)
- **sum**: sum (component {groupingClause}) (in a Component expression within an aggr clause)
Analytic syntax

\[
\text{sum} \left( \text{dataset over (analyticClause)} \right) \quad \text{(in a Data Set expression)}
\]

\[
\text{sum} \left( \text{component over (analyticClause)} \right) \quad \text{(in a Component expression within a calc clause)}
\]

Input parameters

dataset the operand Data Set

component the operand Component

groupingClause see Aggregate invocation

analyticClause see Analytic invocation

Examples of valid syntaxes

See Aggregate and Analytic invocations above, at the beginning of the section.

Semantics for scalar operations

This operator cannot be applied to scalar values.

Input parameters type

dataset :: dataset \{ measure<number> \+_\}

component :: component<number>

Result type

result ::

dataset \{ measure<number> \+_\}

[\ |

component<number>

Additional constraints

None.

Behaviour

The operator returns the sum of the input values.

For other details, see Aggregate and Analytic invocations.

Examples

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>A</td>
<td>XX</td>
<td>3</td>
</tr>
<tr>
<td>2011</td>
<td>A</td>
<td>YY</td>
<td>5</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>YY</td>
<td>7</td>
</tr>
<tr>
<td>2012</td>
<td>A</td>
<td>XX</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>YY</td>
<td>4</td>
</tr>
</tbody>
</table>

Example 1: \[ DS_r := \text{sum} \left( \text{DS_1} \text{ group by } \text{Id}_1 \right) \quad \text{results in:} \]

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>15</td>
</tr>
<tr>
<td>2012</td>
<td>6</td>
</tr>
</tbody>
</table>
Average value : \( \text{avg} \)

**Aggregate syntax**

\[
\text{avg} \left( \text{dataset } \{ \text{groupingClause} \} \right)
\]

*(in a Data Set expression)*

\[
\text{avg} \left( \text{component } \{ \text{groupingClause} \} \right)
\]

*(in a Component expression within an aggr clause)*

**Analytic syntax**

\[
\text{avg} \left( \text{dataset } \text{over} \left( \text{analyticClause} \right) \right)
\]

*(in a Data Set expression)*

\[
\text{avg} \left( \text{component } \text{over} \left( \text{analyticClause} \right) \right)
\]

*(in a Component expression within a calc clause)*

**Input parameters**

- **dataset** the operand Data Set
- **component** the operand Component
- **groupingClause** see Aggregate invocation
- **analyticClause** see Analytic invocation

**Examples of valid syntaxes**

See Aggregate and Analytic invocations above, at the beginning of the section.

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type**

- **dataset** :: dataset \{ measure<number> \_+ \}
- **component** :: component<number>

**Result type**

- **result** :: dataset \{ measure<number> \_+ \}
  | component<number>

**Additional constraints**

None.

**Behaviour**

The operator returns the average of the input values.

For other details, see Aggregate and Analytic invocations.

**Examples**

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>A</td>
<td>XX</td>
<td>3</td>
</tr>
<tr>
<td>2011</td>
<td>A</td>
<td>YY</td>
<td>5</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>YY</td>
<td>7</td>
</tr>
<tr>
<td>2012</td>
<td>A</td>
<td>XX</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>YY</td>
<td>4</td>
</tr>
</tbody>
</table>

**Example 1:**

\[
\text{DS}_r := \text{avg} \left( \text{DS}_1 \text{ group by } \text{Id}_1 \right)
\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>5</td>
</tr>
</tbody>
</table>
Population standard deviation: \texttt{stddev\_pop}

\textbf{Aggregate syntax}

\begin{align*}
\texttt{stddev\_pop} & \ ( \text{dataset} \ \{ \ \text{groupingClause} \ \} ) & \text{(in a Data Set expression)} \\
\texttt{stddev\_pop} & \ ( \text{component} \ \{ \ \text{groupingClause} \ \} ) & \text{(in a Component expression within an \texttt{aggr} clause)}
\end{align*}

\textbf{Analytic syntax}

\begin{align*}
\texttt{stddev\_pop} & \ ( \text{dataset} \ \over\ \{ \ \text{analyticClause} \ \} ) & \text{(in a Data Set expression)} \\
\texttt{stddev\_pop} & \ ( \text{component} \ \over\ \{ \ \text{analyticClause} \ \} ) & \text{(in a Component expression within a \texttt{calc} clause)}
\end{align*}

\textbf{Input parameters}

- \texttt{dataset}: the operand Data Set
- \texttt{component}: the operand Component
- \texttt{groupingClause}: see Aggregate invocation
- \texttt{analyticClause}: see Analytic invocation

\textbf{Examples of valid syntaxes}

See Aggregate and Analytic invocations above, at the beginning of the section.

\textbf{Semantics for scalar operations}

This operator cannot be applied to scalar values.

\textbf{Input parameters type}

- \texttt{dataset} :: dataset \{ measure<number> _+ \}
- \texttt{component} :: component<number>

\textbf{Result type}

- \texttt{result} :: dataset \{ measure<number> _+ \}
  - | component<number>

\textbf{Additional constraints}

None.

\textbf{Behaviour}

The operator returns the "population standard deviation" of the input values.

For other details, see Aggregate and Analytic invocations.

\textbf{Examples}

Given the Data Set \textit{DS\_1}:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>A</td>
<td>XX</td>
<td>3</td>
</tr>
<tr>
<td>2011</td>
<td>A</td>
<td>YY</td>
<td>5</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>YY</td>
<td>7</td>
</tr>
<tr>
<td>2012</td>
<td>A</td>
<td>XX</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>YY</td>
<td>4</td>
</tr>
</tbody>
</table>

Example 1: \( DS\_r := \texttt{stddev\_pop} \ ( \textit{DS\_1} \ \text{group by} \ \text{Id\_1} ) \) results in:
Sample standard deviation: \( \text{stddev\_samp} \)

**Aggregate syntax**

\[
\text{stddev\_samp} \left( \text{dataset} \left( \text{groupingClause} \right) \right) \quad \text{(in a Data Set expression)}
\]

\[
\text{stddev\_samp} \left( \text{component} \left( \text{groupingClause} \right) \right) \quad \text{(in a Component expr. within an aggr clause)}
\]

**Analytic syntax**

\[
\text{stddev\_samp} \left( \text{dataset} \text{ over} \left( \text{analyticClause} \right) \right) \quad \text{(in a Data Set expression)}
\]

\[
\text{stddev\_samp} \left( \text{component} \text{ over} \left( \text{analyticClause} \right) \right) \quad \text{(in a Component expr. within a calc clause)}
\]

**Input parameters**

- **dataset**
  - the operand Data Set
- **component**
  - the operand Component
- **groupingClause**
  - see Aggregate invocation
- **analyticClause**
  - see Analytic invocation

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Examples of valid syntaxes**

See Aggregate and Analytic invocations above, at the beginning of the section.

**Input parameters type**

- **dataset**
  - \( \text{dataset} \left\{ \text{measure<number>}_+ \right\} \)
- **component**
  - \( \text{component<number>} \)

**Result type**

- **result**
  - \( \text{dataset} \left\{ \text{measure<number>}_+ \right\} \)
  - \( \text{component<number>} \)

**Additional constraints**

None.

**Behaviour**

The operator returns the “sample standard deviation” of the input values.

For other details, see Aggregate and Analytic invocations.

**Examples**

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>A</td>
<td>XX</td>
<td>3</td>
</tr>
<tr>
<td>2011</td>
<td>A</td>
<td>YY</td>
<td>5</td>
</tr>
<tr>
<td>2011</td>
<td>B</td>
<td>YY</td>
<td>7</td>
</tr>
</tbody>
</table>
Example 1:  \( DS_r := \text{stddevsamp} \left( DS_1 \text{ group by } \text{Id}_1 \right) \) results in:

<table>
<thead>
<tr>
<th>( \text{Id}_1 )</th>
<th>( \text{Me}_1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>2</td>
</tr>
<tr>
<td>2012</td>
<td>1.4142</td>
</tr>
</tbody>
</table>

Population variance: \( \text{var}_{\text{pop}} \)

**Aggregate syntax**

\[
\text{var}_{\text{pop}} \left( \text{dataset} \{ \text{groupingClause} \} \right) \quad \text{(in a Data Set expression)}
\]

\[
\text{var}_{\text{pop}} \left( \text{component} \{ \text{groupingClause} \} \right) \quad \text{(in a Component expression within an \text{aggr} clause)}
\]

**Analytic syntax**

\[
\text{var}_{\text{pop}} \left( \text{dataset} \over \{ \text{analyticClause} \} \right) \quad \text{(in a Data Set expression)}
\]

\[
\text{var}_{\text{pop}} \left( \text{component} \over \{ \text{analyticClause} \} \right) \quad \text{(in a Component expression within a \text{calc} clause)}
\]

**Input parameters**

- **dataset**: the operand Data Set
- **component**: the operand Component
- **groupingClause**: see Aggregate invocation
- **analyticClause**: see Analytic invocation

**Examples of valid syntaxes**

See Aggregate and Analytic invocations above, at the beginning of the section.

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type**

- **dataset**: \( \text{dataset} \{ \text{measure<number>}_+ \} \)
- **component**: \( \text{component}<\text{number}> \)

**Result type**

- **result**: \( \text{dataset} \{ \text{measure<number>}_+ \} \)

| component<number> |

**Additional constraints**

None.

**Behaviour**

The operator returns the “population variance” of the input values.

For other details, see Aggregate and Analytic invocations.

**Examples**

Given the Data Set \( DS_1 \):

\[
DS_1
\]

<table>
<thead>
<tr>
<th>2012</th>
<th>A</th>
<th>XX</th>
<th>2</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>YY</td>
<td>4</td>
</tr>
</tbody>
</table>
Example 1: \[ DS_r := \text{var} \_ \text{pop} ( DS \_1 \text{ group by } \mathrm{id} \_1 ) \] results in:

\[
\begin{array}{c|c}
\hline
\text{id} \_1 & \text{Me} \_1 \\
2011 & 2,6667 \\
2012 & 1 \\
\hline
\end{array}
\]

Sample variance: \text{var} \_ \text{samp}

Aggregate syntax
\[
\text{var} \_ \text{samp} ( \text{dataset} \{ \text{grouping} \text{Clause} \} ) \quad (\text{in a Data Set expression})
\]
\[
\text{var} \_ \text{samp} ( \text{component} \{ \text{grouping} \text{Clause} \} ) \quad (\text{in a Component expression within an aggr clause})
\]

Analytic syntax
\[
\text{var} \_ \text{samp} ( \text{dataset} \over ( \text{analytic} \text{Clause} ) ) \quad (\text{in a Data Set expression})
\]
\[
\text{var} \_ \text{samp} ( \text{component} \over ( \text{analytic} \text{Clause} ) ) \quad (\text{in a Component expression within a calc clause})
\]

Input parameters
- dataset: the operand Data Set
- component: the operand Component
- groupingClause: see Aggregate invocation
- analyticClause: see Analytic invocation

Examples of valid syntaxes
See Aggregate and Analytic invocations above, at the beginning of the section.

Semantics for scalar operations
This operator cannot be applied to scalar values.

Input parameters type
- dataset: \text{dataset} \{ \text{measure}<\text{number}>_+ \}
- component: \text{component}<\text{number}>

Result type
- result: \text{dataset} \{ \text{measure}<\text{number}>_+ \}
- \text{component}<\text{number}>

Additional constraints
None.

Behaviour
The operator returns the sample variance of the input values.
For other details, see Aggregate and Analytic invocations.
Examples

Given the Data Set \( DS_1 \)

\[
\begin{array}{cccc}
Id_1 & Id_2 & Id_3 & Me_1 \\
2011 & A & XX & 3 \\
2011 & A & YY & 5 \\
2011 & B & YY & 7 \\
2012 & A & XX & 2 \\
2012 & B & YY & 4 \\
\end{array}
\]

Example 1: \( DS_r := \text{varsamp} \ ( DS_1 \ \text{group by} \ Id_1) \)  
results in:

\[
\begin{array}{c}
Id_1 & Me_1 \\
2011 & 4 \\
2012 & 2 \\
\end{array}
\]

First value:  \( \text{first\_value} \)

Syntax

\[
\text{first\_value} \ (\ \text{dataset} \ \text{over} \ (\ \text{analyticClause}) \ ) \quad \text{(in a Data Set expression)}
\]

\[
\text{first\_value} \ (\ \text{component} \ \text{over} \ (\ \text{analyticClause}) \ ) \quad \text{(in a Component expression within a \text{calc} clause)}
\]

Input parameters

dataset the operand Data Set

component the operand Component

analyticClause see Analytic invocation

Examples of valid syntaxes

See Analytic invocation above, at the beginning of the section.

Semantics for scalar operations

This operator cannot be applied to scalar values.

Input parameters type

dataset :: dataset \{ measure<scalar> \_+ \}

component :: component<scalar>

Result type

result :: dataset  

| component<scalar>

Additional constraints

The Aggregate invocation is not allowed.

Behaviour

The operator returns the first value (in the value order) of the set of Data Points that belong to the same analytic window as the current Data Point.
When invoked at Data Set level, it returns the first value for each Measure of the input Data Set. The first value of different Measures can result from different Data Points.

When invoked at Component level, it returns the first value of the specified Component.

For other details, see Analytic invocation.

Examples

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>1993</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1994</td>
<td>4</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1995</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1996</td>
<td>6</td>
<td>8</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1993</td>
<td>9</td>
<td>3</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1994</td>
<td>5</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1995</td>
<td>10</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1996</td>
<td>2</td>
<td>7</td>
</tr>
</tbody>
</table>

Example 1:

DS_r := first_value ( DS_1 over ( partition by  Id_1, Id_2 order by  Id_3 data points between 1 preceding and 1 following) ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>1993</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1994</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1995</td>
<td>4</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1996</td>
<td>6</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1993</td>
<td>5</td>
<td>3</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1994</td>
<td>5</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1995</td>
<td>2</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1996</td>
<td>2</td>
<td>2</td>
</tr>
</tbody>
</table>

Last value : last_value

Syntax

last_value ( dataset over ( analyticClause ) )

(last in a Data Set expression)

last_value ( component over ( analyticClause ) )

(last in a Component expression within a calc clause)

Input parameters

dataset the operand Data Set

component the operand Component

analyticClause see Analytic invocation
Examples of valid syntaxes
See Analytic invocation above, at the beginning of the section.

Semantics for scalar operations
This operator cannot be applied to scalar values.

Input parameters type
\[
\text{dataset ::= dataset \{measure<scalar> \_+\}}
\]
\[
\text{component :: component<scalar>}
\]

Result type
\[
\text{result ::= dataset | component<scalar>}
\]

Additional constraints
The Aggregate invocation is not allowed.

Behaviour
The operator returns the last value (in the value order) of the set of Data Points that belong to the same analytic window as the current Data Point.

When invoked at Data Set level, it returns the last value for each Measure of the input Data Set. The last value of different Measures can result from different Data Points.

When invoked at Component level, it returns the last value of the specified Component.

For other details, see Analytic invocation.

Examples
Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>1993</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1994</td>
<td>4</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1995</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1996</td>
<td>6</td>
<td>8</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1993</td>
<td>9</td>
<td>3</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1994</td>
<td>5</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1995</td>
<td>10</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1996</td>
<td>2</td>
<td>7</td>
</tr>
</tbody>
</table>

Example 1:
\[
\text{DS_r := last_value ( DS_1 over ( partition by Id_1, Id_2 order by Id_3 data points between 1 preceding and 1 following ) )}
\]

results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>1993</td>
<td>4</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1994</td>
<td>7</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1995</td>
<td>7</td>
<td>9</td>
</tr>
</tbody>
</table>
Lag :  lag

Syntax

in a Data Set expression:
\[
\text{lag ( dataset \{ offset \{ defaultValue \} \} over \{ partitionClause \} orderClause} )
\]

In a Component expression within a calc clause:
\[
\text{lag ( component \{ offset \{ defaultValue \} \} over \{ partitionClause \} orderClause} )
\]

Input parameters

- dataset :: dataset
- component :: component
- offset :: integer [ value > 0 ]
- defaultValue :: scalar

Result type

- result :: dataset | component

Examples of valid syntaxes

See Analytic invocation above, at the beginning of the section.

Semantics for scalar operations

This operator cannot be applied to scalar values.

Input parameters type

- dataset :: dataset
- component :: component
- offset :: integer [ value > 0 ]
- defaultValue :: scalar

Result type

- result :: dataset | component

Additional constraints

- The Aggregate invocation is not allowed.
- The windowClause of the Analytic invocation syntax is not allowed.

Behaviour

In the ordered set of Data Points of the current partition, the operator returns the value(s) taken from the Data Point at the specified physical offset prior to the current Data Point.

If defaultValue is not specified then the value returned when the offset goes outside the partition is NULL.

For other details, see Analytic invocation.

Examples

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>DS_1</th>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>1996</td>
<td>7</td>
<td>8</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1993</td>
<td>9</td>
<td>4</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1994</td>
<td>10</td>
<td>4</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1995</td>
<td>10</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1996</td>
<td>10</td>
<td>7</td>
<td></td>
</tr>
</tbody>
</table>
Example 1: $DS_r := \text{lag} ( DS_1 , 1 \text{ over ( partition by Id_1 , Id_2 order by Id_3 ) } )$

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>1993</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1994</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1995</td>
<td>4</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1996</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1993</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1994</td>
<td>9</td>
<td>3</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1995</td>
<td>5</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1996</td>
<td>10</td>
<td>2</td>
</tr>
</tbody>
</table>

lead : lead

Syntax

in a Data Set expression:

$\text{lead} ( \text{dataset}, \{ \text{offset}, \text{defaultValue} \} ) \text{ over ( } \{ \text{partitionClause} \} \text{ orderClause } ) )$

in a Component expression within a calc clause:

$\text{lead} ( \text{component}, \{ \text{offset}, \text{defaultValue} \} ) \text{ over ( } \{ \text{partitionClause} \} \text{ orderClause } ) )$

Input parameters

dataset :: dataset
component :: component
offset :: the relative position beyond the current Data Point
defaultValue :: the value returned when the offset goes outside the partition.
partitionClause :: see Analytic invocation
orderClause :: see Analytic invocation

Examples of valid syntaxes

See Analytic invocation above, at the beginning of the section.

Semantics for scalar operations

This operator cannot be applied to scalar values.
offset :: integer [ value > 0 ]
default value :: scalar

Result type
result :: dataset | component

Additional constraints
The Aggregate invocation is not allowed.
The windowClause of the Analytic invocation syntax is not allowed.

Behaviour
In the ordered set of Data Points of the current partition, the operator returns the value(s) taken from the Data Point at the specified physical offset beyond the current Data Point.
If defaultValue is not specified, then the value returned when the offset goes outside the partition is NULL.
For other details, see Analytic invocation.

Examples
Given the Data Set DS_1

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>1993</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1994</td>
<td>4</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1995</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1996</td>
<td>6</td>
<td>8</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1993</td>
<td>9</td>
<td>3</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1994</td>
<td>5</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1995</td>
<td>10</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1996</td>
<td>2</td>
<td>7</td>
</tr>
</tbody>
</table>

Example 1: DS_r := lead ( DS_1 , 1 over ( partition by Id_1 , Id_2 order by Id_3 ) ) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>1993</td>
<td>4</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1994</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1995</td>
<td>6</td>
<td>8</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>1996</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1993</td>
<td>5</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1994</td>
<td>10</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1995</td>
<td>2</td>
<td>7</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>1996</td>
<td>NULL</td>
<td>NULL</td>
</tr>
</tbody>
</table>

Rank:

Syntax
rank ( over ( { partitionClause } orderClause ) ) (in a Component expression within a calc clause)
**Input parameters**

- **partitionClause** see Analytic invocation
- **orderClause** see Analytic invocation

**Examples of valid syntaxes**

See Analytic invocation above, at the beginning of the section.

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type**

- **dataset** :: dataset
- **component** :: component

**Result type**

- **result** ::
  - dataset  { measure<integer> int_var }
  - | component<integer>

**Additional constraints**

- The invocation at Data Set level is not allowed.
- The Aggregate invocation is not allowed.
- The **windowClause** of the Analytic invocation syntax is not allowed.

**Behaviour**

The operator returns an order number (rank) for each Data Point, starting from the number 1 and following the order specified in the **orderClause**. If some Data Points are in the same order according to the specified **orderClause**, the same order number (rank) is assigned and a gap appears in the sequence of the assigned ranks (for example, if four Data Points have the same rank 5, the following assigned rank would be 9).

For other details, see Analytic invocation.

**Examples**

Given the Data Set **DS_1**:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>2000</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2001</td>
<td>4</td>
<td>9</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2002</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2003</td>
<td>6</td>
<td>8</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2000</td>
<td>9</td>
<td>3</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2001</td>
<td>5</td>
<td>4</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2002</td>
<td>10</td>
<td>2</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2003</td>
<td>5</td>
<td>7</td>
</tr>
</tbody>
</table>

**Example 1:**

```
DS_r := DS_1 [ calc Me2 := rank ( over ( partition by Id_1, Id_2 order by Me_1 ) ) ]
```

**DS_r**

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>2000</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2001</td>
<td>4</td>
<td>2</td>
</tr>
</tbody>
</table>
Ratio to report: \( \text{ratio\_to\_report} \)

**Syntax**

\[
\text{ratio\_to\_report} \left( \text{dataset \ over ( partitionClause )} \right) \quad \text{(in a Data Set expression)}
\]

\[
\text{ratio\_to\_report} \left( \text{component \ over ( partitionClause )} \right) \quad \text{(in a Component expr. within a calc clause)}
\]

**Input parameters**

- **dataset**: the operand Data Set
- **component**: the operand Component
- **partitionClause**: see Analytic invocation

**Examples of valid syntaxes**

See Analytic invocation above, at the beginning of the section.

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type**

- **dataset::**
  - dataset \{ measure\text{number}\_+ \}
- **component::**
  - component\text{number}

**Result type**

- **result::**
  - dataset \{ measure\text{number}\_+ \}
  - component\text{number}

**Additional constraints**

- The Aggregate invocation is not allowed.
- The \text{orderClause} and \text{windowClause} of the Analytic invocation syntax are not allowed.

**Behaviour**

The operator returns the ratio between the value of the current Data Point and the sum of the values of the partition which the current Data Point belongs to.

**Examples**

Given the Data Set \text{DS}_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>2000</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2001</td>
<td>4</td>
<td>3</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2002</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2003</td>
<td>6</td>
<td>1</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2000</td>
<td>12</td>
<td>0</td>
</tr>
</tbody>
</table>
Example 1: \[ DS_r := \frac{\text{ratio_to_report}(DS_1 \text{ over (partition by } \text{Id}_1, \text{Id}_2))}{...} \] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>XX</td>
<td>2000</td>
<td>0.15</td>
<td>0.1</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2001</td>
<td>0.2</td>
<td>0.3</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2002</td>
<td>0.35</td>
<td>0.5</td>
</tr>
<tr>
<td>A</td>
<td>XX</td>
<td>2003</td>
<td>0.3</td>
<td>0.1</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2000</td>
<td>0.3</td>
<td>0</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2001</td>
<td>0.2</td>
<td>0.8</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2002</td>
<td>0.15</td>
<td>0.5</td>
</tr>
<tr>
<td>A</td>
<td>YY</td>
<td>2003</td>
<td>0.35</td>
<td>-0.3</td>
</tr>
</tbody>
</table>
check_datapoint

Syntax

\[
\text{check_datapoint ( op, dpr \{ components listComp \} \{ output \})}
\]

\[
\begin{align*}
\text{listComp} & \ := \ \text{comp \{ , comp \} }^* \\
\text{output} & \ := \ \text{invalid | all | all_measures}
\end{align*}
\]

Input parameters

\begin{itemize}
\item \text{op} \quad \text{the Data Set to check}
\item \text{dpr} \quad \text{the Data Point Ruleset to be used}
\item \text{listComp} \quad \text{if dpr is defined on Value Domains then listComp is the list of Components of op to be associated (in positional order) to the conditioning Value Domains defined in dpr. If dpr is defined on Variables then listComp is the list of Components of op to be associated (in positional order) to the conditioning Variables defined in dpr (for documentation purposes).}
\item \text{comp} \quad \text{Component of \text{op}}
\item \text{output} \quad \text{specifies the Data Points and the Measures of the resulting Data Set:}
\begin{itemize}
\item \text{invalid} \quad \text{the resulting Data Set contains a Data Point for each Data Point of \text{op} and each Rule in \text{dpr} that evaluates to FALSE on that Data Point. The resulting Data Set has the Measures of \text{op}}.
\item \text{all} \quad \text{the resulting Data Set contains a data point for each Data Point of \text{op} and each Rule in \text{dpr}. The resulting Data Set has the \text{boolean Measure bool_var}.}
\item \text{all_measures} \quad \text{the resulting Data Set contains a Data Point for each Data Point of \text{op} and each Rule in \text{dpr}. The resulting dataset has the Measures of \text{op} and the \text{boolean Measure bool_var}.}
\end{itemize}
\end{itemize}

If not specified then \text{output} is assumed to be \text{invalid}. See the Behaviour for further details.

Examples of valid syntaxes

\begin{itemize}
\item \text{check_datapoint ( DS1, DPR invalid )}
\item \text{check_datapoint ( DS1, DPR all_measures )}
\end{itemize}

Semantics for scalar operations

This operator cannot be applied to scalar values.

Input parameters type:

\begin{itemize}
\item \text{op} :: dataset
\item \text{dpr} :: name < datapoint >
\item \text{comp} :: name < component >
\end{itemize}

Result type:

\begin{itemize}
\item \text{result} :: dataset
\end{itemize}

Additional constraints

\begin{itemize}
\item If \text{dpr} is defined on Value Domains then it is mandatory to specify \text{listComp}. The Components specified in \text{listComp} must belong to the operand \text{op} and be defined on the Value Domains specified in the signature of \text{dpr}.
\item If \text{dpr} is defined on Variables then the Components specified in the signature of \text{dpr} must belong to the operand \text{op}.
\item If \text{dpr} is defined on Variables and \text{listComp} is specified then the Components specified in \text{listComp} are the same, in the same order, as those specified in \text{op} (they are provided for documentation purposes).
\end{itemize}
Behaviour

It returns a Data Set having the following Components:

- the Identifier Components of op
- the Identifier Component ruleid whose aim is to identify the Rule that has generated the actual Data Point (it contains at least the Rule name specified in dpr)
- if the output parameter is invalid: the original Measures of op (no boolean measure)
- if the output parameter is all: the boolean Measure bool_var whose value is the result of the evaluation of a rule on a Data Point (TRUE, FALSE or NULL).
- if the output parameter is all_measures: the original measures of op and the boolean Measure bool_var whose value is the result of the evaluation of a rule on a Data Point (TRUE, FALSE or NULL).
- the Measure errorcode that contains the errorcode specified in the rule
- the Measure errorlevel that contains the errorlevel specified in the rule

A Data Point of op can produce several Data Points in the resulting Data Set, each of them with a different value of ruleid. If output is invalid then the resulting Data Set contains a Data Point for each Data Point of op and each rule of dpr that evaluates to FALSE. If output is all or all_measures then the resulting Data Set contains a Data Point for each Data Point of op and each rule of dpr.

Examples

define datapoint ruleset dpr1 ( variable Id_3, Me_1 ) is
  when Id_3 = “CREDIT” then Me_1 >= 0 errorcode “Bad credit”
  ; when Id_3 = “DEBIT” then Me_1 >= 0 errorcode “Bad debit”
end datapoint ruleset

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>I</td>
<td>CREDIT</td>
<td>10</td>
</tr>
<tr>
<td>2011</td>
<td>I</td>
<td>DEBIT</td>
<td>-2</td>
</tr>
<tr>
<td>2012</td>
<td>I</td>
<td>CREDIT</td>
<td>10</td>
</tr>
<tr>
<td>2012</td>
<td>I</td>
<td>DEBIT</td>
<td>2</td>
</tr>
</tbody>
</table>

DS_r := check_datapoint ( DS_1, dpr1 )

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>ruleid</th>
<th>obs_value</th>
<th>errorcode</th>
<th>errorlevel</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>I</td>
<td>DEBIT</td>
<td>dpr1_2</td>
<td>-2</td>
<td>Bad debit</td>
<td></td>
</tr>
</tbody>
</table>

DS_r := check_datapoint ( DS_1, dpr1 all )

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>ruleid</th>
<th>bool_var</th>
<th>errorcode</th>
<th>errorlevel</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>I</td>
<td>CREDIT</td>
<td>dpr1_1</td>
<td>true</td>
<td></td>
<td></td>
</tr>
<tr>
<td>2011</td>
<td>I</td>
<td>CREDIT</td>
<td>dpr1_2</td>
<td>true</td>
<td></td>
<td></td>
</tr>
<tr>
<td>2011</td>
<td>I</td>
<td>DEBIT</td>
<td>dpr1_1</td>
<td>true</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

8 The content of ruleid maybe personalised in the implementation
<table>
<thead>
<tr>
<th>Year</th>
<th>Type</th>
<th>Component</th>
<th>Condition</th>
<th>Result</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>DEBIT</td>
<td>dpr1_2</td>
<td>false</td>
<td>Bad debit</td>
</tr>
<tr>
<td>2012</td>
<td>CREDIT</td>
<td>dpr1_1</td>
<td>true</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>CREDIT</td>
<td>dpr1_2</td>
<td>true</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>DEBIT</td>
<td>dpr1_1</td>
<td>true</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>DEBIT</td>
<td>dpr1_2</td>
<td>true</td>
<td></td>
</tr>
</tbody>
</table>

**check_hierarchy**

**Syntax**

```plaintext
check_hierarchy ( op, hr { condition condComp { condComp}* } { rule ruleComp } { mode } { input } { output })
```

- **mode** ::= non_null | non_zero | partial_null | partial_zero | always_null | always_zero
- **input** ::= dataset | dataset_priority
- **output** ::= invalid | all | all_measures

**Input parameters**

- **op** the Data Set to be checked
- **hr** the hierarchical Ruleset to be used
- **condComp** condComp is a Component of op to be associated (in positional order) to the conditioning Value Domains or Variables defined in hr (if any).
- **ruleComp** ruleComp is the Identifier Component of op to be associated to the rule Value Domain or Variable defined in hr.
- **mode** this parameter specifies how to treat the possible missing Data Points corresponding to the Code Items in the left and right sides of the rules and which Data Points are produced in output. The meaning of the possible values of the parameter is explained below.
- **input** this parameter specifies the source of the values used as input of the comparisons. The meaning of the possible values of the parameter is explained below.
- **output** this parameter specifies the structure and the content of the resulting dataset. The meaning of the possible values of the parameter is explained below.

**Examples of valid syntaxes**

- `check_hierarchy ( DS1, HR_2 non_null dataset invalid )`
- `check_hierarchy ( DS1, HR_3 non_zero dataset_priority all )`

**Input parameters type**

- **op** ::= dataset { measure<number> _ }  
- **hr** ::= name < hierarchical >
- **condComp** ::= name < component >
- **ruleComp** ::= name < identifier >

**Result type**

- **result** ::= dataset { measure<number> _ }  

**Additional constraints**

If hr is defined on Value Domains then it is mandatory to specify the condition (if any in the ruleset hr) and the rule parameters. Moreover, the Components specified as condComp and ruleComp must belong to the operand...
op and must take values on the Value Domains corresponding, in positional order, to the ones specified in the
condition and rule parameter of hr.

If hr is defined on Variables, the specification of condComp and ruleComp is not needed, but they can be
specified all the same if it is desired to show explicitly in the invocation which are the involved Components: in
this case, the condComp and ruleComp must be the same and in the same order as the Variables specified in in
the condition and rule signatures of hr.

**Behaviour**

The **check_hierachy** operator applies the Rules of the Ruleset hr to check the Code Items Relations between
the Code Items present in op (as for the Code Items Relations, see the User Manual - section "Generic Model for
Variables and Value Domains"). The operator checks if the relation between the left and the right member is
fulfilled, giving TRUE in positive case and FALSE in negative case.

The Attribute propagation rule is applied on each group of Data Points which contributes to the same Data Point
of the result.

The behaviours relevant to the different options of the input parameters are the following.

First, the parameter input is used to determine the source of the Data Points used as input of the
check_hierachy. The possible options of the parameter input and the corresponding behaviours are the
following:

- **dataset**: this option addresses the case where all the input Data Points of all the Rules of the Ruleset are
  expected to be taken from the input Data Set (the operand op).
  For each Rule of the Ruleset and for each item on the left and right sides of the Rule, the
  operator takes the input Data Points exclusively from the operand op.

- **dataset_priority**: this option addresses the case where the input Data Points of all the Rules of the Ruleset are
  preferably taken from the input Data Set (the operand op), however if a valid Measure value
  for an expected Data Point is not found in op, the attempt is made to take it from the computed
  output of a (possible) other Rule.
  For each Rule of the Ruleset and for each item on the left and right sides of the Rule:
  - if the item is not defined as the result (left side) of another Rule that applies the Code Item
    relation "is equal to" (=), the current Rule takes the input Data Points from the operand op.
  - if the item is defined as result of another Rule \( R \) that applies the Code Item relation "is
    equal to" (=), then:
    - if an expected input Data Point exists in op and its Measure is not NULL, then the
      current Rule takes such Data Point from op;
    - if an expected input Data Point does not exist in op or its measure is NULL, then
      the current Rule takes the Data Point (if any) that has the same Identifiers' values
      from the computed output of the other Rule \( R \);

If the parameter input is not specified then it is assumed to be **dataset**.

Then the parameter **mode** is considered, to determine the behaviour for missing Data Points and for the Data
Points to be produced in the output. The possible options of the parameter **mode** and the corresponding
behaviours are the following:

- **non_null**: the result Data Point is produced when all the items involved in the comparison exist and have
  not NULL Measure value (i.e., when no Data Point corresponding to the Code Items of the left
  and right sides of the rule is missing or has NULL Measure value); under this option, in
  evaluating the comparison, the possible missing Data Points corresponding to the Code Items
  of the left and right sides of the rule are considered existing and having a NULL Measure value;

- **non_zero**: the result Data Point is produced when at least one of the items involved in the comparison
  exist and have Measure not equal to 0 (zero); the possible missing Data Points corresponding
  to the Code Items of the left and right sides of the rule are considered existing and having a
  Measure value equal to 0;

- **partial_null**: the result Data Point is produced if at least one Data Point corresponding to the Code Items of
  the left and right sides of the rule is found (whichever is its Measure value); the possible
  missing Data Points corresponding to the Code Items of the left and right sides of the rule are
  considered existing and having a NULL Measure value;
the result Data Point is produced if at least one Data Point corresponding to the Code Items of
the left and right sides of the rule is found (whichever is its Measure value); the possible
missing Data Points corresponding to the Code Items of the left and right sides of the rule are
considered existing and having a Measure value equal to 0 (zero);

the result Data Point is produced in any case; the possible missing Data Points corresponding
to the Code Items of the left and right sides of the rule are considered existing and having a
Measure value equal to NULL;

the result Data Point is produced in any case; the possible missing Data Points corresponding
to the Code Items of the left and right sides of the rule are considered existing and having a
Measure value equal to 0 (zero);

If the parameter mode is not specified, then it is assumed to be **non_null**.

The following table summarizes the behaviour of the options of the parameter “mode”

<table>
<thead>
<tr>
<th>OPTION of the MODE PARAMETER:</th>
<th>Missing Data Points are considered:</th>
<th>Null Data Points are considered:</th>
<th>Condition for evaluating the rule</th>
<th>Returned Data Points</th>
</tr>
</thead>
<tbody>
<tr>
<td>Non_null</td>
<td>NULL</td>
<td>NULL</td>
<td>If all the involved Data Points are not NULL</td>
<td>Only not NULL Data Points (Zeros are returned too)</td>
</tr>
<tr>
<td>Non_zero</td>
<td>Zero</td>
<td>NULL</td>
<td>If at least one of the involved Data Points is &lt;&gt; zero</td>
<td>Only not zero Data Points (NULLS are returned too)</td>
</tr>
<tr>
<td>Partial_null</td>
<td>NULL</td>
<td>NULL</td>
<td>If at least one of the involved Data Points is not NULL</td>
<td>Data Points of any value (NULL, not NULL and zero too)</td>
</tr>
<tr>
<td>Partial_zero</td>
<td>Zero</td>
<td>NULL</td>
<td>If at least one of the involved Data Points is not NULL</td>
<td>Data Points of any value (NULL, not NULL and zero too)</td>
</tr>
<tr>
<td>Always_null</td>
<td>NULL</td>
<td>NULL</td>
<td>Always</td>
<td>Data Points of any value (NULL, not NULL and zero too)</td>
</tr>
<tr>
<td>Always_zero</td>
<td>Zero</td>
<td>NULL</td>
<td>Always</td>
<td>Data Points of any value (NULL, not NULL and zero too)</td>
</tr>
</tbody>
</table>

Finally the parameter output is considered, to determine the structure and content of the resulting Data Set. The possible options of the parameter output and the corresponding behaviours are the following:

**all** all the Data Points produced by the comparison are returned, both the valid ones (TRUE) and the invalid ones (FALSE) besides the possible NULL ones. The result of the comparison is returned in the boolean Measure bool_var. The original Measure Component of the Data Set op is not returned.

**invalid** only the invalid (FALSE) Data Points produced by the comparison are returned. The result of the comparison (boolean Measure bool_var) is not returned. The original Measure Component of the Data Set op is returned and contains the Measure values taken from the Data Points on the left side of the rule.

**all_measures** all the Data Points produced by the comparison are returned, both the valid ones (TRUE) and the invalid ones (FALSE) besides the possible NULL ones. The result of the comparison is returned in the boolean Measure bool_var. The original Measure Component of the Data Set op is returned and contains the Measure values taken from the Data Points on the left side of the rule.

If the parameter output is not specified then it is assumed to be **invalid**.
In conclusion, the operator returns a Data Set having the following Components:

- all the Identifier Components of op
- the additional Identifier Component ruleid, whose aim is to identify the Rule that has generated the actual Data Point (it contains at least the Rule name specified in hr 
  
if the output parameter is all: the \textit{boolean} Measure \texttt{bool\_var} whose values are the result of the evaluation of the Rules (TRUE, FALSE or NULL).
- if the output parameter is invalid: the original Measure of op, whose values are taken from the Measure values of the Data Points of the left side of the Rule
- if the output parameter is all\_measures: the \textit{boolean} Measure \texttt{bool\_var}, whose value is the result of the evaluation of a Rule on a Data Point (TRUE, FALSE or NULL), and the original Measure of op, whose values are taken from the Measure values of the Data Points of the left side of the Rule
- the Measure imbalance, which contains the difference between the Measure values of the Data Points on the left side of the Rule and the Measure values of the corresponding calculated Data Points on the right side of the Rule
- the Measure errorcode, which contains the \texttt{errorcode} value specified in the Rule
- the Measure errorlevel, which contains the \texttt{errorlevel} value specified in the Rule

Note that a generic Data Point of \textit{op} can produce several Data Points in the resulting Data Set, one for each Rule in which the Data Point appears as the left member of the comparison.

\textbf{Examples}

See also the examples in \texttt{define hierarchic ruleset}.

Given the following hierarchical ruleset:

\begin{verbatim}
define hierarchical ruleset HR_1 (valuedomain rule VD_1) is
    R010 : A = J + K + L  ; errorlevel 5
    ; R020 : B = M + N + O  ; errorlevel 5
    ; R030 : C = P + Q  ; errorcode XX  ; errorlevel 5
    ; R040 : D = R + S  ; errorlevel 1
    ; R060 : F = Y + W + Z  ; errorlevel 7
    ; R070 : G = B + C
    ; R080 : H = D + E  ; errorlevel 0
    ; R090 : I = D + G  ; errorcode YY  ; errorlevel 0
    ; R100 : M >= N  ; errorlevel 5
    ; R110 : M <= G  ; errorlevel 5
end hierarchical ruleset
\end{verbatim}

\begin{verbatim}
And given \texttt{the} operand Data Set DS_1 (where At_1 is viral and the propagation rule says that the alphabetic order prevails the NULL prevails on the alphabetic characters and the Attribute value for missing Data Points is assumed as NULL):
\end{verbatim}

\begin{verbatim}
DS_1
\end{verbatim}

\begin{verbatim}
\begin{tabular}{|c|c|c|}
\hline
Id_1 & Id_2 & Me_1 \\
\hline
2010 & A & 5 \\
2010 & B & 11 \\
2010 & C & 0 \\
2010 & G & 19 \\
2010 & H & NULL \\
2010 & I & 14 \\
2010 & M & 2 \\
\hline
\end{tabular}
\end{verbatim}

\textit{The content of ruleid maybe personalised in the implementation}
Example 1: \[ DS_r := \text{check}_\text{hierarchy} ( DS_1, \text{HR}_1 \text{ rule Id}_2 \text{ partial_null all}) \] results in:

<table>
<thead>
<tr>
<th>\text{DS}_r</th>
<th>\text{Id}_1</th>
<th>\text{Id}_2</th>
<th>\text{ruleid}</th>
<th>\text{Bool}_\text{var}</th>
<th>\text{imbalance}</th>
<th>\text{errorcode}</th>
<th>\text{errorlevel}</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>N</td>
<td>5</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>O</td>
<td>4</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>P</td>
<td>7</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>Q</td>
<td>-7</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>S</td>
<td>3</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>T</td>
<td>9</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>U</td>
<td>NULL</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>V</td>
<td>6</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

\[ \text{check} \]

**Syntax**

\[ \text{check} ( \text{op} \{ \text{errorcode} \text{ errorcode} \} \{ \text{errorlevel} \text{ errorlevel} \} \{ \text{imbalance} \text{ imbalance} \} \{ \text{output} \}) \]

\[ \text{output} ::= \text{invalid} | \text{all} \]

**Input parameters**

- \text{op} a boolean Data Set (a boolean condition expressed on one or more Data Sets)
- \text{errorcode} the error code to be produced when the condition evaluates to FALSE. It must be a valid value of the errorcode\_vd Value Domain (or string if the errorcode\_vd Value Domain is not found).
- \text{errorlevel} the error level to be produced when the condition evaluates to FALSE. It must be a valid value of the errorlevel\_vd Value Domain (or integer if the errorcode\_vd Value Domain is not found).
- \text{imbalance} the imbalance to be computed. imbalance is a numeric mono-measure Data Set with the same Identifiers of \text{op}. If not specified then \text{imbalance} is NULL.
- \text{output} specifies which Data Points are returned in the resulting Data Set:
invalid returns the Data Points of op for which the condition evaluates to FALSE.

all returns all Data Points of op

If not specified then output is all.

Examples of valid syntaxes

check ( DS1 > DS2 errorcode myerrorcode errorlevel myerrorlevel imbalance DS1 - DS2 invalid )

Input parameters type:

op :: dataset
errorcode :: errorcode_vd
errorlevel :: errorlevel_vd
imbalance :: number

Result type:

result :: dataset

Additional constraints

op has exactly a boolean Measure Component.

Behaviour

It returns a Data Set having the following components:

- the Identifier Components of op
- a boolean Measure named bool_var that contains the result of the evaluation of the boolean dataset op
- the Measure imbalance that contains the specified imbalance
- the Measure errorcode that contains the specified errorcode
- the Measure errorlevel that contains the specified errorlevel

If output is all then all data points are returned. If output is invalid then only the Data Points where bool_var is FALSE are returned.

Examples

Given the Data Sets DS_1 and DS_2:

<table>
<thead>
<tr>
<th>ID_2</th>
<th>ID_3</th>
<th>ME_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>D</td>
<td>25</td>
</tr>
<tr>
<td>2011</td>
<td>D</td>
<td>35</td>
</tr>
<tr>
<td>2012</td>
<td>D</td>
<td>45</td>
</tr>
<tr>
<td>2013</td>
<td>D</td>
<td>55</td>
</tr>
<tr>
<td>2014</td>
<td>D</td>
<td>50</td>
</tr>
</tbody>
</table>
Example 1:  

\[
\text{DS}_r := \text{check ( DS1} \geq \text{DS2 imbalance DS1 - DS2 })
\]

returns:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>bool_var</th>
<th>imbalance</th>
<th>errorcode</th>
<th>errorlevel</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>I</td>
<td>FALSE</td>
<td>-8</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2011</td>
<td>I</td>
<td>TRUE</td>
<td>0</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2012</td>
<td>I</td>
<td>TRUE</td>
<td>0</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2013</td>
<td>I</td>
<td>FALSE</td>
<td>-3</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2014</td>
<td>I</td>
<td>TRUE</td>
<td>0</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2015</td>
<td>I</td>
<td>TRUE</td>
<td>0</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2010</td>
<td>D</td>
<td>FALSE</td>
<td>-25</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2011</td>
<td>D</td>
<td>TRUE</td>
<td>0</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2012</td>
<td>D</td>
<td>TRUE</td>
<td>5</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2013</td>
<td>D</td>
<td>TRUE</td>
<td>0</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2014</td>
<td>D</td>
<td>FALSE</td>
<td>-15</td>
<td>NULL</td>
<td>NULL</td>
</tr>
<tr>
<td>2015</td>
<td>D</td>
<td>TRUE</td>
<td>0</td>
<td>NULL</td>
<td>NULL</td>
</tr>
</tbody>
</table>
if-then-else :

Syntax
if condition then thenOperand else elseOperand

Input parameters
condition :: a Boolean condition (dataset, component or scalar)
thenOperand :: the operand returned when condition evaluates to true
elseOperand :: the operand returned when condition evaluates to false

Examples of valid syntaxes
if A > B then A else B

Semantics for scalar operations
The if operator returns thenOperand if condition evaluates to true, elseOperand otherwise. For example,
considering the statement:
if x1 > x2 then 2 else 5,
for x1 = 3, x2 = 0 it returns 2
for x1 = 0, x2 = 3 it returns 5

Input Parameters type
condition :: | component < boolean > |
| boolean |
thenOperand :: | component |
| scalar |
elseOperand :: | component |
| scalar |

Result type
result :: dataset |
| component |
| scalar |

Additional constraints
- The operands thenOperand and elseOperand must be of the same scalar type.
- If the operation is at scalar level, thenOperand and elseOperand are scalar then condition must be scalar too (a boolean scalar).
- If the operation is at Component level, at least one of thenOperand and elseOperand is a Component (the other one can be scalar) and condition must be a Component too (a boolean Component); thenOperand, elseOperand and the other Components referenced in condition must belong to the same Data Set.
- If the operation is at Data Set level, at least one of thenOperand and elseOperand is a Data Set (the other one can be scalar) and condition must be a Data Set too (having a unique boolean Measure) and must have the same Identifiers as thenOperand or/and ElseOperand
  - If thenOperand and elseOperand are both Data Sets then they must have the same Components in the same roles
  - If one of thenOperand and elseOperand is a Data Set and the other one is a scalar, the Measures of the operand Data Set must be all of the same scalar type as the scalar operand.
For operations at Component level, the operation is applied for each Data Point of the unique input Data Set, the if-then-else operator returns the value from the thenOperand Component when condition evaluates to true, otherwise it returns the value from the elseOperand Component. If one of the operands thenOperand or elseOperand is scalar, such a scalar value can be returned depending on the outcome of the condition.

For operations at Data Set level, the if-then-else operator returns the Data Point from thenOperand when the Data Point of condition having the same Identifiers’ values evaluates to true, and returns the Data Point from elseOperand otherwise. If one of the operands thenOperand or elseOperand is scalar, such a scalar value can be returned (depending on the outcome of the condition) and in this case it feeds the values of all the Measures of the result Data Point.

The behaviour for two Data Sets can be procedurally explained as follows. First the condition Data Set is evaluated, then its true Data Points are inner joined with thenOperand and its false Data Points are inner joined with elseOperand, finally the union is made of these two partial results (the condition ensures that there cannot be conflicts in the union).

**Examples**

**Example 1:** given the operand Data Sets DS_cond, DS_1, DS_2:

<table>
<thead>
<tr>
<th>DS_cond</th>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>M</td>
<td>5451780</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>F</td>
<td>5643070</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>M</td>
<td>5449803</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>F</td>
<td>5673231</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>M</td>
<td>23099012</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>F</td>
<td>23719207</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>M</td>
<td>31616281</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>F</td>
<td>33671580</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>I</td>
<td>Total</td>
<td>M</td>
<td>28726599</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>I</td>
<td>Total</td>
<td>F</td>
<td>30667608</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>A</td>
<td>Total</td>
<td>M</td>
<td>NULL</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>A</td>
<td>Total</td>
<td>F</td>
<td>NULL</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>DS_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>DS_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id_1</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
<tr>
<td>2012</td>
</tr>
</tbody>
</table>
DS_r := if ( DS_cond#4 = "F" ) then DS_1 else DS_2

returns:

<table>
<thead>
<tr>
<th>DS_r</th>
<th>id_1</th>
<th>id_2</th>
<th>id_3</th>
<th>id_4</th>
<th>me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>F</td>
<td>25.8</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>F</td>
<td>NULL</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>I</td>
<td>Total</td>
<td>F</td>
<td>20.9</td>
<td></td>
</tr>
</tbody>
</table>

Nvl : nvl

Syntax
nvl ( op1 , op2 )

Input parameters
- op1: the first operand
- op2: the second operand

Examples of valid syntaxes
- nvl ( ds1#m1, 0 )

Semantics for scalar operations
The operator nvl returns op2 when op1 is null, otherwise op1. For example:
- nvl ( 5, 0 ) returns 5
- nvl ( null, 0 ) returns 0

Input Parameters type
- op1 ::
  | dataset
  | component<scalar>
  | scalar
- op2 ::
  | dataset
  | component
  | <scalar>

Result type
- result ::
  | dataset
  | component
  | scalar

Additional constraints
- If op1 and op2 are scalar values then they must be of the same type.
- If op1 and op2 are Components then they must be of the same type.
- If op1 and op2 are Data Sets then they must have the same Components.

Behaviour
The operator nvl returns the value from op2 when the value from op1 is null, otherwise it returns the value from op1.
The operator has the typical behaviour of the operators applicable on two scalar values or Data Sets or Data Set Components.
Also the following statement gives the same result: if isnull ( op1 ) then op2 else op1

Examples
Example 1: Given the input Data Set DS_1

Example 2: Given the input Data Set DS_2
### DS_1

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>11094850</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>11123034</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>417546</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>5401267</td>
</tr>
<tr>
<td>2012</td>
<td>N</td>
<td>Total</td>
<td>Total</td>
<td>NULL</td>
</tr>
</tbody>
</table>

$$DS_r := nvl(DS_1, 0)$$ returns:

### DS_r

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Id_4</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2012</td>
<td>B</td>
<td>Total</td>
<td>Total</td>
<td>11094850</td>
</tr>
<tr>
<td>2012</td>
<td>G</td>
<td>Total</td>
<td>Total</td>
<td>11123034</td>
</tr>
<tr>
<td>2012</td>
<td>S</td>
<td>Total</td>
<td>Total</td>
<td>0</td>
</tr>
<tr>
<td>2012</td>
<td>M</td>
<td>Total</td>
<td>Total</td>
<td>417546</td>
</tr>
<tr>
<td>2012</td>
<td>F</td>
<td>Total</td>
<td>Total</td>
<td>5401267</td>
</tr>
<tr>
<td>2012</td>
<td>N</td>
<td>Total</td>
<td>Total</td>
<td>0</td>
</tr>
</tbody>
</table>
Filtering Data Points: filter

**Syntax**

```
op [ filter filterCondition ]
```

**Input parameters**

- `op`: the operand
- `filterCondition`: the filter condition

**Examples of valid syntaxes**

- `DS_1 [ filter Me_3 > 0 ]`
- `DS_1 [ filter Me_3 + Me_2 <= 0 ]`

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type:**

- `op`: dataset
- `filterCondition`: component<boolean>

**Result type:**

- result: dataset

**Additional constraints:**

- None.

**Behaviour**

The operator takes as input a Data Set (`op`) and a boolean Component expression (`filterCondition`) and filters the input Data Points according to the evaluation of the condition. When the expression is TRUE the Data Point is kept in the result, otherwise it is not kept (in other words, it filters out the Data Points of the operand Data Set for which `filterCondition` condition evaluates to FALSE or NULL).

**Examples**

Given the Data Set DS_1:

<p>| | | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>XX</td>
<td>2</td>
<td>E</td>
</tr>
<tr>
<td>1</td>
<td>A</td>
<td>YY</td>
<td>2</td>
<td>F</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>XX</td>
<td>20</td>
<td>F</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>YY</td>
<td>1</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>XX</td>
<td>4</td>
<td>E</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>YY</td>
<td>9</td>
<td>F</td>
</tr>
</tbody>
</table>

**Example 1:**

`DS_r := DS_1 [ filter Id_1 = 1 and Me_1 < 10 ]` results in:

<p>| | | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Calculation of a Component: \( \text{calc} \)

**Syntax**

\[
\text{op} \{ \text{calc} \{ \text{calcRole} \} \text{calcComp} := \text{calcExpr} \{ , \{ \text{calcRole} \} \text{calcComp} := \text{calcExpr} \}^* \}
\]

\[
\text{calcRole} ::= \text{identifier} | \text{measure} | \text{attribute} | \text{viral attribute}
\]

**Input parameters**

- **op**: the operand
- **calcRole**: the role to be assigned to a Component to be calculated
- **calcComp**: the name of a Component to be calculated
- **calcExpr**: expression at component level, having only Components of the input Data Sets as operands, used to calculate a Component

**Examples of valid syntaxes**

- \( \text{DS}_1 \{ \text{calc Me}_3 := \text{Me}_1 + \text{Me}_2 \} \)

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type:**

- **op**: dataset
- **calcComp**: name < component >
- **calcExpr**: component<scalar>

**Result type:**

- **result**: dataset

**Additional constraints**

The **calcComp** parameter cannot be the name of an Identifier component.

All the components used in **calcComp** must belong to the operand Data Set **op**.

**Behaviour**

The operator calculates new Identifier, Measure or Attribute Components on the basis of sub-expressions at Component level. Each Component is calculated through an independent sub-expression. It is possible to specify the role of the calculated Component among **measure**, **identifier**, **attribute**, or **viral attribute**, therefore the **calc** clause can be used also to change the role of a Component when possible. The keyword **viral** allows controlling the virality of the calculated Attributes (for the attribute propagation rule see the User Manual). When the role is omitted, the following rule is applied: if the component exists in the operand Data Set then it maintains its role; if the component does not exist in the operand Data Set then its role is Measure.

The **calcExpr** sub-expressions are independent one another, they can only reference Components of the input Data Set and cannot use Components generated, for example, by other **calcExpr**. If the calculated Component is a new Component, it is added to the output Data Set. If the Calculated component is a Measure or an Attribute that already exists in the input Data Set, the calculated values overwrite the original values. If the calculated Component is an Identifier that already exists in the input Data Set, an exception is raised because overwriting an Identifier Component is forbidden for preserving the functional behaviour. Analytic invocations can be used in the **calc** clause.

**Examples**

- Given the Data Set **DS_1**:
Example 1: \( DS_r := DS_1 \{ \text{calc } Me_1 := Me_1 \times 2 \} \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>CA</td>
<td>40</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>CA</td>
<td>4</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>CA</td>
<td>4</td>
</tr>
</tbody>
</table>

Example 2: \( DS_r := DS_1 \{ \text{calc attribute } At_1 := \text{"EP"} \} \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>CA</td>
<td>40</td>
<td>EP</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>CA</td>
<td>4</td>
<td>EP</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>CA</td>
<td>4</td>
<td>EP</td>
</tr>
</tbody>
</table>

Aggregation: \( \text{aggr} \)

**Syntax**

\[
\text{op} \left[ \text{aggr aggrClause} \{ \text{groupingClause} \} \right]
\]

- \( \text{aggrClause} := \{ \text{aggrRole} \} \)
- \( \text{aggrComp} := \text{aggrExpr} \)
- \( \{, \{ \text{aggrRole} \} \} \)
- \( \text{groupingClause} := \{ \text{by} \} \)
- \( \text{group except} \)
- \( \text{group all} \)
- \( \text{having} \)

**Input Parameters**

- \( \text{op} \) the operand
- \( \text{aggrClause} \) clause that specifies the required aggregations, i.e., the aggregated Components to be calculated, their roles and their calculation algorithm, to be applied on the joined and filtered Data Points
- \( \text{aggrRole} \) the role of the aggregated Component to be calculated
- \( \text{aggrComp} \) the name of the aggregated Component to be calculated; this is a dependent Component of the result (Measure or Attribute, not Identifier)
aggrExpr

Expression at component level, having only Components of the input Data Sets as operands, which invokes an aggregate operator (e.g. `avg`, `count`, `max` ... , see also the corresponding sections) to perform the desired aggregation. Note that the `count` operator is used in an `aggrClause` without parameters, e.g:

```
DS_1 [ aggr Me_1 := count () group by Id_1 ]
```

groupingClause

The following alternative grouping options:

- **group by**
  - The Data Points are grouped by the values of the specified Identifiers.
  - (groupingId). The Identifiers not specified are dropped in the result.

- **group except**
  - The Data Points are grouped by the values of the Identifiers not specified as groupingId. The Identifiers specified as groupingId are dropped in the result.

- **group all**
  - Converts the values of an Identifier Component using conversionExpr and keeps all the resulting Identifiers.

groupingId

Identifier Component to be kept (in the `group by` clause) or dropped (in the `group except` clause).

conversionExpr

Specifies a conversion operator (e.g., `time_agg`) to convert an Identifier from finer to coarser granularity. The conversion operator is applied on an Identifier of the operand Data Set op.

havingCondition

A condition (boolean expression) at component level, having only Components of the input Data Sets as operands (and possibly constants), to be fulfilled by the groups of Data Points: only groups for which havingCondition evaluates to `TRUE` appear in the result. The havingCondition refers to the groups specified through the groupingClause, therefore it must invoke aggregate operators (e.g. `avg`, `count`, `max` ... , see also the section Aggregate invocation). A correct example of havingCondition is:

````
max(obs_value) < 1000
```

Instead the condition `obs_value < 1000` is not a right havingCondition, because it refers to the values of the single Data Points and not to the groups. The `count` operator is used in a havingCondition without parameters, e.g:

```
sum (DS_1 group by id1 having count () >= 10 )
```

Examples of valid syntaxes

- `DS_1 [ aggr M1 := min ( Me_1 ) group by Id_1, Id_2 ]`
- `DS_1 [ aggr M1 := min ( Me_1 ) group except Id_1, Id_2 ]`

Semantics for scalar operations

This operator cannot be applied to scalar values.

Input parameters type:

- `op :: dataset`
- `aggrComp :: name < component >`
- `aggrExpr :: component<scalar>`
- `groupingId :: name <identifier>`
- `conversionExpr :: identifier<scalar>`
- `havingCondition :: component<boolean>`

Result type:

- `result :: dataset`

Additional constraints

- The `aggrComp` parameter cannot be the name of an Identifier component.
- All the components used in `aggrExpr` must belong to the operand Data Set op.
- The `conversionExpr` parameter applies just one conversion operator to just one Identifier belonging to the input Data Set. The basic scalar type of the Identifier must be compatible with the basic scalar type of the conversion operator.
The operator \textbf{aggr} calculates aggregations of dependent Components (Measures or Attributes) on the basis of sub-expressions at Component level. Each Component is calculated through an independent sub-expression. It is possible to specify the role of the calculated Component among \textit{measure attribute}, or \textit{viral attribute}. The substring \textbf{viral} allows to control the virality of Attributes, if the Attribute propagation rule is adopted (see the \textit{User Manual}). When the role is omitted, the following rule is applied: if the component exists in the operand Data Set then it maintains its role; if the component does not exist in the operand Data Set then its role is Measure.

The \textit{aggrExpr} sub-expressions are independent of one another, they can only reference Components of the input Data Set and cannot use Components generated, for example, by other \textit{aggrExpr} sub-expressions. The \textbf{aggr} computed Measures and Attributes are the only Measures and Attributes returned in the output Data Set (plus the possible viral Attributes). The sub-expressions must contain only Aggregate operators, which are able to compute an aggregated Value relevant to a group of Data Points. The groups of Data Points to be aggregated are specified through the \textit{groupingClause}, which allows the following alternative options.

- \textbf{group by} the Data Points are grouped by the values of the specified Identifiers. The Identifiers not specified are dropped in the result.
- \textbf{group except} the Data Points are grouped by the values of the Identifiers not specified in the clause. The specified Identifiers are dropped in the result.
- \textbf{group all} converts an Identifier Component using \textit{conversionExpr} and keeps all the other Identifiers.

The \textbf{having} clause is used to filter groups in the result by means of an aggregate condition evaluated on the single groups (for example the minimum number of Data Points in the group). If no grouping clause is specified, then all the input Data Points are aggregated in a single group and the clause returns a Data Set that contains a single Data Point and has no Identifiers. The Attributes calculated through the \textbf{aggr} clauses are maintained in the result. For all the other Attributes that are defined as \textit{viral}, the Attribute propagation rule is applied (for the semantics, see the Attribute Propagation Rule section in the \textit{User Manual}).

\textbf{Examples}

Given the Data Set DS_1:

\begin{verbatim}
<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>XX</td>
<td>0</td>
</tr>
<tr>
<td>1</td>
<td>A</td>
<td>YY</td>
<td>2</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>XX</td>
<td>3</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>YY</td>
<td>5</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>XX</td>
<td>7</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>YY</td>
<td>2</td>
</tr>
</tbody>
</table>
\end{verbatim}

\textit{Example1:} \quad DS_r := DS_1 \ [ \text{aggr Me_1:= sum( Me_1 ) group by Id_1 , Id_2 } ] \quad \text{results in:}

\begin{verbatim}
<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>2</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>8</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>9</td>
</tr>
</tbody>
</table>
\end{verbatim}

\textit{Example2:} \quad DS_r := DS_1 \ [ \text{aggr Me_3:= min( Me_1 ) group except Id_3 } ] \quad \text{results in:}

\begin{verbatim}
<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Me_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>2</td>
</tr>
</tbody>
</table>

\end{verbatim}
Example 3:

\[
\text{DS}_r := \text{DS}_1 \left[ \text{aggr } \text{Me}_1 := \text{sum( Me}_1 \text{) }, \text{Me}_2 := \text{max( Me}_1 \text{)} \right]
\text{group by } \text{Id}_1, \text{Id}_2
\text{having } \text{avg( Me}_1 \text{)} > 2 \]

results in:

<table>
<thead>
<tr>
<th>\text{Id}_1</th>
<th>\text{Id}_2</th>
<th>\text{Me}_1</th>
<th>\text{Me}_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>8</td>
<td>5</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>9</td>
<td>7</td>
</tr>
</tbody>
</table>

Maintaining Components: \text{keep}

**Syntax**

\[
\text{op [ keep comp } \{, \text{ comp } \}^* ]
\]

**Input parameters**

- \text{op} the operand
- \text{comp} a component to keep

**Examples of valid syntaxes**

- \text{DS}_1 \left[ \text{keep Me}_2, \text{Me}_3 \right]

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type:**

- \text{op} :: \text{dataset}
- \text{comp} :: \text{name} < \text{component}>

**Result type:**

- \text{result} :: \text{dataset}

**Additional constraints:**

- All the Components \text{comp} must belong to the input Data Set \text{op}.
- The Components \text{comp} cannot be Identifiers in \text{op}.

**Behaviour**

The operator takes as input a Data Set (\text{op}) and some Component names of such a Data Set (\text{comp}). These Components can be Measures or Attributes of \text{op} but not Identifiers. The operator maintains the specified Components, drops all the other dependent Components of the Data Set (Measures and Attributes) and maintains the independent Components (Identifiers) unchanged. This operation corresponds to a projection in the usual relational join semantics (specifying which columns will be projected in among Measures and Attributes).

**Examples**

Given the Data Set \text{DS}_1:
Example 1: \( DS_r := DS_1 [\text{keep } Me_1] \) results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>Me_2</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>2010</td>
<td>A</td>
<td>XX</td>
<td>20</td>
<td>36</td>
<td>E</td>
</tr>
<tr>
<td>2010</td>
<td>A</td>
<td>YY</td>
<td>4</td>
<td>9</td>
<td>F</td>
</tr>
<tr>
<td>2010</td>
<td>B</td>
<td>XX</td>
<td>9</td>
<td>10</td>
<td>F</td>
</tr>
</tbody>
</table>

**Removal of Components:** \( \text{drop} \)

**Syntax**

\[ \text{op } [\text{drop } \text{comp} \{ , \text{comp} \}] \]

**Input parameters**

- \( \text{op} \): the operand
- \( \text{comp} \): a Component to drop

**Examples of valid syntaxes**

- \( DS_1 [\text{drop } Me_2, Me_3] \)

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input parameters type:**

- \( \text{op}:: \) dataset
- \( \text{comp}:: \) name \(<\text{component}>\)

**Result type:**

- \( \text{result}:: \) dataset

**Additional constraints:**

- All the Components \( \text{comp} \) must belong to the input Data Set \( \text{op} \).
- The Components \( \text{comp} \) cannot be Identifiers in \( \text{op} \).

**Behaviour**

The operator takes as input a Data Set \( \text{op} \) and some Component names of such a Data Set \( \text{comp} \). These Components can be Measures or Attributes of \( \text{op} \) but not Identifiers. The operator drops the specified Components and maintains all the other Components of the Data Set. This operation corresponds to a projection in the usual relational join semantics (specifying which columns will be projected out).

**Examples**

Given the Data Set \( DS_1 \):
**Example 1:**

\[
\text{DS}_r := \text{DS}_1 \ [ \text{drop } \text{At}_1 ]
\]

results in:

\[
\begin{array}{cccc}
\text{Id}_1 & \text{Id}_2 & \text{Id}_3 & \text{Me}_1 \\
2010 & A & \text{XX} & 20 \\
2010 & A & \text{YY} & 4 \\
2010 & B & \text{XX} & 9 \\
\end{array}
\]

**Change of Component name:** rename

**Syntax**

\[
op [ \text{rename } \text{comp}_\text{from} \text{ to } \text{comp}_\text{to} \{ , \text{comp}_\text{from} \text{ to } \text{comp}_\text{to} \}^* ]
\]

**Input Parameters**

- \text{op} \quad \text{the operand}
- \text{comp}_\text{from} \quad \text{the original name of the Component to rename}
- \text{comp}_\text{to} \quad \text{the new name of the Component after the renaming}

**Examples of valid syntaxes**

- \text{DS}_1 \ [ \text{rename Me}_2 \text{ to Me}_3 ]

**Semantics for scalar operations**

This operator cannot be applied to scalar values.

**Input Parameters type**

- \text{op}:: \quad \text{dataset}
- \text{comp}_\text{from}:: \quad \text{name} < \text{component}>
- \text{comp}_\text{to}:: \quad \text{name} < \text{component}>

**Result type**

- \text{result}:: \quad \text{dataset}

**Additional constraints**

The corresponding pairs of Components before and after the renaming (\text{dsc}_\text{from} and \text{dsc}_\text{to}) must be defined on the same Value Domain and the same Value Domain Subset.

The components used in \text{dsc}_\text{from} must belong to the input Data Set and the component used in the \text{dsc}_\text{to} cannot have the same names as other Components of the result Data Set.

**Behaviour**

The operator assigns new names to one or more Components (Identifier, Measure or Attribute Components).

The resulting Data Set, after renaming the specified Components, must have unique names of all its Components (otherwise a runtime error is raised). Only the Component name is changed and not the Component Values, therefore the new Component must be defined on the same Value Domain and Value Domain Subset as the original Component (see also the IM in the User Manual). If the name of a Component defined on a different Value Domain or Set is assigned, an error is raised. In other words, rename is a transformation of the variable without any change in its values.
Examples

Given the Data Set DS_1:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>B</td>
<td>XX</td>
<td>20</td>
<td>F</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>YY</td>
<td>1</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>XX</td>
<td>4</td>
<td>E</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>YY</td>
<td>9</td>
<td>F</td>
</tr>
</tbody>
</table>

Example1: DS_r := DS_1 [ rename Me_1 to Me_2, At_1 to At_2] results in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_2</th>
<th>At_2</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>B</td>
<td>XX</td>
<td>20</td>
<td>F</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>YY</td>
<td>1</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>XX</td>
<td>4</td>
<td>E</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>YY</td>
<td>9</td>
<td>F</td>
</tr>
</tbody>
</table>

Pivoting: pivot

Syntax

\[ \text{op [ pivot identifier, measure ]} \]

Input parameters

- \text{op}: the operand
- \text{identifier}: the Identifier Component of \text{op} to pivot
- \text{measure}: the Measure Component of \text{op} to pivot

Examples of valid syntaxes

- \text{DS_1 [ pivot Id_2, Me_1 ]}

Semantics for scalar operations

- This operator cannot be applied to scalar values.

Input Parameters type

- \text{op}: \text{dataset}
- \text{identifier}: \text{name < identifier >}
- \text{measure}: \text{name < measure >}

Result type

- \text{result}: \text{dataset}

Additional constraints

- The Measures created by the operator according to the behaviour described below must be defined on the same Value Domain as the input Measure.

Behaviour
The operator transposes several Data Points of the operand Data Set into a single Data Point of the resulting Data Set. The semantics of **pivot** can be procedurally described as follows.

1. It creates a virtual Data Set $VDS$ as a copy of $op$.
2. It drops the Identifier Component $identifier$ and all the Measure Components from $VDS$.
3. It groups $VDS$ by the values of the remaining Identifiers.
4. For each distinct value of $identifier$ in $op$, it adds a corresponding measure to $VDS$, named as the value of $identifier$. These Measures are initialized with the NULL value.
5. For each Data Point of $op$, it finds the Data Point of $VDS$ having the same values as for the common Identifiers and assigns the value of $measure$ (taken from the current Data Point of $op$) to the Measure of $VDS$ having the same name as the value of $identifier$ (taken from the Data Point of $op$).

The result of the last step is the output of the operation.

Note that **pivot** may create Measures whose names are non-regular (i.e. they may contain special characters, reserved keywords, etc.) according to the rules about the artefact names described in the User Manual (see the section “The artefact names” in the chapter “VTL Transformations”). As said in the User Manual, those names must be quoted to be referenced within an expression.

**Examples**

Given the Data Set $DS_1$:

<table>
<thead>
<tr>
<th>$Id_1$</th>
<th>$Id_2$</th>
<th>$Me_1$</th>
<th>$At_1$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>5</td>
<td>E</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>2</td>
<td>F</td>
</tr>
<tr>
<td>1</td>
<td>C</td>
<td>7</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>3</td>
<td>E</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>4</td>
<td>E</td>
</tr>
<tr>
<td>2</td>
<td>C</td>
<td>9</td>
<td>F</td>
</tr>
</tbody>
</table>

**Example1:** $DS_r := Ds_1 \ [ \text{pivot } Id_2, \ Me_1 ]$ results in:

<table>
<thead>
<tr>
<th>$Id_1$</th>
<th>A</th>
<th>B</th>
<th>C</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>5</td>
<td>2</td>
<td>7</td>
</tr>
<tr>
<td>2</td>
<td>3</td>
<td>4</td>
<td>9</td>
</tr>
</tbody>
</table>

**Unpivoting:** **unpivot**

**Syntax**

$op \ [ \text{unpivot } identifier, \ measure ]$

**Input parameters**

- $op$: the dataset operand
- $identifier$: the Identifier Component to be created
- $measure$: the Measure Component to be created

**Examples of valid syntaxes**
Semantics for scalar operations

This operator cannot be applied to scalar values.

Input Parameters type

\[ \text{op} :: \text{dataset} \]
\[ \text{identifier} :: \text{name} < \text{identifier} > \]
\[ \text{measure} :: \text{name} < \text{measure} > \]

Result type

\[ \text{result} :: \text{dataset} \]

Additional constraints

All the measures of \( \text{op} \) must be defined on the same Value Domain.

Behaviour

The \text{unpivot} operator transposes a single Data Point of the operand Data Set into several Data Points of the result Data set. Its semantics can be procedurally described as follows.

1. It creates a virtual Data Set VDS as a copy of \( \text{op} \)
2. It adds the Identifier Component \( \text{identifier} \) and the Measure Component \( \text{measure} \) to VDS.
3. For each Data Point \( \text{DP} \) and for each Measure \( \text{M} \) of \( \text{op} \) whose value is not NULL, the operator inserts a Data Point into VDS whose values are assigned as specified in the following points
4. The VDS Identifiers other than \( \text{identifier} \) are assigned the same values as the corresponding Identifiers of the \( \text{op} \) Data Point
5. The VDS \( \text{identifier} \) is assigned a value equal to the \text{name} of the Measure \( M \) of \( \text{op} \)
6. The VDS \( \text{measure} \) is assigned a value equal to the \text{value} of the Measure \( M \) of \( \text{op} \)

The result of the last step is the output of the operation.

When a Measure is NULL then \text{unpivot} does not create a Data Point for that Measure.

Note that in general pivoting and unpivoting are not exactly symmetric operations, i.e., in some cases the unpivot operation applied to the pivoted Data Set does not recreate exactly the original Data Set (before pivoting).

Examples

Given the Data Set \( \text{DS}_1 \):

\[
\begin{array}{c|c|c|c}
\text{Id}_1 & \text{A} & \text{B} & \text{C} \\
\hline
1 & 5 & 2 & 7 \\
2 & 3 & 4 & 9 \\
\end{array}
\]

Example1: \( \text{DS}_r := \text{DS}_1 \ [ \text{unpivot Id}_2, \text{Me}_1] \) results in:

\[
\begin{array}{c|c|c}
\text{Id}_1 & \text{Id}_2 & \text{Me}_1 \\
\hline
1 & A & 5 \\
1 & B & 2 \\
1 & C & 7 \\
2 & A & 3 \\
\end{array}
\]
Subspace : sub

Syntax

do [ sub identifier = value { , identifier = value }* ]

Input parameters

op :: dataset
identifier :: Identifier Component of the input Data Set
value :: valid value for identifier

Examples of valid syntaxes

DS_r := DS_1 [ Id_2 = "A", Id_5 = 1 ]

Semantics for scalar operations

This operator cannot be applied to scalar values.

Input Parameters type

op :: dataset
identifier :: name < identifier >
value :: scalar

Result type

result :: dataset

Additional constraints

The specified Identifier Components identifier(s) must belong to the input Data Set op.
Each Identifier Component can be specified only once.
The specified value must be an allowed value for identifier.

Behaviour

The operator returns a Data Set in a subspace of the one of the input Dataset. Its behaviour can be procedurally described as follows:

1. It creates a virtual Data Set VDS as a copy of op
2. It maintains the Data Points of VDS for which identifier = value (for all the specified identifier) and eliminates all the Data Points for which identifier <> value (even for only one specified identifier)
3. It projects out ("drops", in VTL terms) all the identifier(s)

The result of the last step is the output of the operation.

The resulting Data Set has the Identifier Components that are not specified as identifier(s) and has the same Measure and Attribute Components of the input Data Set.

The result Data Set does not violate the functional constraint because after the filter of the step 2, all the remaining identifier(s) do not contain the same Values for all the Data Points. In other words, given that the input Data Set is a 1st order function and therefore does not contain duplicates, the result Data Set is a 1st order function as well. To show this, let K_1, K_n be the Identifier components for the generic input Data Set DS.

Let us suppose that K_1, K_n are assigned to fixed values by using the subspace operator. A duplicate could arise only if in the result there are two Data Points DP_a and DP_b having the same value for K_{m+1},..,K_n, but this is impossible since such Data Points had same K_1,..,K_m in the original Data Set DS, which did not contain duplicates.
If we consider the vector space of Data Points individuated by the n-uples of Identifier components of a Data Set $\text{DS}(K_1,\ldots,K_n)$ (along, e.g., with the operators of sum and multiplication), we have that the subspace operator actually performs a subsetting of such space into another space with fewer Identifiers. This can be also seen as the equivalent of a dice operation performed on hyper-cubes in multi-dimensional data warehousing.

**Examples**

Given the Data Set $\text{DS}_1$:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_2</th>
<th>Id_3</th>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>A</td>
<td>XX</td>
<td>20</td>
<td>F</td>
</tr>
<tr>
<td>1</td>
<td>A</td>
<td>YY</td>
<td>1</td>
<td>F</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>XX</td>
<td>4</td>
<td>E</td>
</tr>
<tr>
<td>1</td>
<td>B</td>
<td>YY</td>
<td>9</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>XX</td>
<td>7</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>A</td>
<td>YY</td>
<td>5</td>
<td>E</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>XX</td>
<td>12</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>B</td>
<td>YY</td>
<td>15</td>
<td>F</td>
</tr>
</tbody>
</table>

**Example 1:**  $\text{DS}_r := \text{DS}_1 \mid \text{sub} \ Id_1 = 1, \ Id_2 = “A”$  
result in:

<table>
<thead>
<tr>
<th>Id_3</th>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>XX</td>
<td>20</td>
<td>F</td>
</tr>
<tr>
<td>YY</td>
<td>1</td>
<td>F</td>
</tr>
</tbody>
</table>

**Example 2:**  $\text{DS}_r := \text{DS}_1 \mid \text{sub} \ Id_1 = 1, \ Id_2 = “B”, \ Id_3 = “YY”$  
result in:

<table>
<thead>
<tr>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>9</td>
<td>F</td>
</tr>
</tbody>
</table>

**Example 3:**  $\text{DS}_r := \text{DS}_1 \mid \text{sub} \ Id_2 = “A” + \text{DS}_1 \mid \text{sub} \ Id_2 = “B”$  
result in:

<table>
<thead>
<tr>
<th>Id_1</th>
<th>Id_3</th>
<th>Me_1</th>
<th>At_1</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>XX</td>
<td>24</td>
<td>F</td>
</tr>
<tr>
<td>1</td>
<td>YY</td>
<td>10</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>XX</td>
<td>19</td>
<td>F</td>
</tr>
<tr>
<td>2</td>
<td>YY</td>
<td>20</td>
<td>F</td>
</tr>
</tbody>
</table>

Assuming that $\text{At}_1$ is viral and that in the propagation rule the greater value prevails, result in: