

# Package ‘rLDCP’

May 9, 2026

**Type** Package

**Title** Text Generation from Data

**Version** 1.0.2

**Date** 2017-11-10

**Author** Patricia Conde-Clemente [aut, cre], Jose M. Alonso [aut], Gracian Trivino [aut]

**Maintainer** Patricia Conde-Clemente <patricia.condeclemente@gmail.com>

**Description** Linguistic Descriptions of Complex Phenomena (LDCP) is an architecture and methodology that allows us to model complex phenomena, interpreting input data, and generating automatic text reports customized to the user needs (see <[doi:10.1016/j.ins.2016.11.002](https://doi.org/10.1016/j.ins.2016.11.002)> and <[doi:10.1007/s00500-016-2430-5](https://doi.org/10.1007/s00500-016-2430-5)>). The proposed package contains a set of methods that facilitates the development of LDCP systems. Its main goal is increasing the visibility and practical use of this research line.

**License** GPL (>= 2) | file LICENSE

**URL** <http://phedes.com/rLDCP>

**LazyData** FALSE

**NeedsCompilation** no

**RoxygenNote** 6.0.1

**Suggests** testthat

**Imports** XML (>= 3.98-1.4), methods

**Repository** CRAN

**Date/Publication** 2017-11-10 16:42:11 UTC

## Contents

cp . . . . .	2
data_structure . . . . .	3
degree_mf . . . . .	4
fuzzy_partitions . . . . .	5
fuzzy_rule . . . . .	5

fuzzy_rules . . . . .	6
generate_code . . . . .	7
glmp . . . . .	7
infer_rules . . . . .	8
ldcp . . . . .	9
ldcp_run . . . . .	10
operator . . . . .	10
pm . . . . .	11
pm_infer . . . . .	13
pm_multidimensional . . . . .	14
pm_report . . . . .	15
report_template . . . . .	16
trapezoid_mf . . . . .	17
triangle_mf . . . . .	17
validate_xml . . . . .	18
xml2rldcp . . . . .	19

<b>Index</b>	<b>20</b>
--------------	-----------

---

cp	<i>Define the CP</i>
----	----------------------

---

### Description

In general, CP corresponds with specific parts of the analyzed phenomenon at a certain degree of granularity. To create a computational model of the analyzed phenomenon, the designer analyzes the everyday use of natural language about the monitored phenomenon with the aim of identifying different parts (units of information or granules) based on his/her subjective perceptions. According with Zadeh (1996), a granule is a clump of elements which are drawn together by indistinguishability, similarity, proximity or functionality. The GLMP handles granules by using CPs.

### Usage

```
cp(name, a, b = NULL, r = NULL)
```

### Arguments

name	is the identifier of the CP.
a	is a vector $A = (a_1, a_2, \dots, a_n)$ of linguistic expressions that represents the whole linguistic domain of CP, e.g. we have the linguistic domain "statistical data" that is represented with three linguistic variables (bad, good, very good).
b	is a vector $B = (b_1, b_2, \dots, b_n)$ of linguistic expressions (words or sentences in natural language) that represents the reliability of the CP, e.g., the reliability of the "statistical data" are (low, moderate, high). By default ( $b = \text{NULL}$ ), the CP does not manage information about reliability.

`r` is a vector  $R = (r_1, r_2, \dots, r_n)$  of relevance degrees  $0 \leq r_i \leq 1$  assigned to each  $a_i$  in the specific context, e.g., the relevance of the linguistic expressions (bad, good, very good) is (0.5, 0.5, 1) means the perception of "very good" is more relevant than the other two choices. By default ( $r = \text{NULL}$ ), the function create a  $r$  vector with the maximum degree of relevance for all linguistic expression, e.g., (1,1,1).

### Value

The generated  $\text{CP} = \text{list}(a, w, r, b, wb)$  where  $w$  and  $wb$  are vectors with the validity degrees ( $w_i$  and  $wb_i$  in  $[0,1]$ ) of the linguistic expressions in  $a$  and  $b$  respectively. These vectors are initialized with 0.

### Examples

```
myCP <- cp("myCP", c("bad", "good", "very good"))
myCP <- cp("myCP", c("bad", "good", "very good"), c("low", "moderate", "high"))
myCP <- cp("myCP", c("bad", "good", "very good"), r=c(1,0.8,0.9))
myCP <- cp("myCP", c("bad", "good", "very good"), c("low", "moderate", "high"), c(1,0.8,0.9))
```

---

data_structure	<i>Define the data structure</i>
----------------	----------------------------------

---

### Description

Data structure provides the GLMP input. Its constructor receives the input values and the method that defines the data structure, i.e., the set of preprocessing techniques.

### Usage

```
data_structure(input, method)
```

### Arguments

`input` is the input data. May be a vector, list or matrix with numbers.

`method` is the function with the data preprocessing techniques needed to prepare the GLMP input. The method must have one argument, the input data:

```
my_method <- function(input)
```

### Value

The generated  $\text{data\_structure} = \text{list}(input, method)$

**Examples**

```

values <- matrix(c(34,11,9,32), ncol=2)

my_method <- function (input){
  output <- c(mean(input[,1]), mean(input[,2]))
  output
}
my_data_structure <- data_structure(values,my_method)

```

---

degree\_mf

*Define generic calculation of fuzzy membership degrees*


---

**Description**

It is a generic function in charge of computing fuzzy membership degrees. Namely, it identifies the specific membership function to consider and run the related method for computing the membership degree for a given input value. It takes as input an object (trapezoid\_mf, triangle\_mf and fuzzy\_partitions) and the related input values

**Usage**

```
degree_mf(shape, input)
```

**Arguments**

shape	is the object (trapezoid_mf, triangle_mf and fuzzy_partitions) to dispatch to.
input	is the value to be assess.

**Value**

the membership degree for a given input values.

**Examples**

```

w <- degree_mf(triangle_mf(450,450,550),450)

w <- degree_mf(fuzzy_partitions(triangle_mf(450,450,550),
                                triangle_mf(450,550,600),
                                trapezoid_mf(550,600,800,1000),
                                triangle_mf(800,1000,1300),
                                trapezoid_mf(1000,1300,1500,1500)),450)

```

---

fuzzy\_partitions      *Define the fuzzy partitions*

---

### Description

It is a constructor of fuzzy partitions, it defines a set of membership functions. It takes as input a set of trapezoid\_mf or triangle\_mf or objects in the shape\_mf class.

### Usage

```
fuzzy_partitions(...)
```

### Arguments

...                    are the different partitions, e.g., trapezoid\_mf or triangle\_mf.

### Value

the (fuzzy\_partitions <- list(...))

### Examples

```
fuzzy_partitions(triangle_mf(450,450,550),
                 triangle_mf(450,550,600),
                 trapezoid_mf(550,600,800, 1000),
                 triangle_mf(800,1000,1300),
                 trapezoid_mf(1000,1300,1500,1500))
```

---

fuzzy\_rule              *Define the fuzzy rule*

---

### Description

We define a fuzzy rule using the numbers 1 and 0. rule(0,0,1,0,0, 0,0,1,0,0, 0,0,1,0,0, 0,0,1)

This is an example of fuzzy\_rule(0,0,1,0,0,1). In the fuzzy rule the number 1 means that the linguistic expression is included and the number 0 means that the linguistic expression is not included.

### Usage

```
fuzzy_rule(...)
```

### Arguments

...                    the 0 and 1 that compose the fuzzy rule.

**Value**

```
the_fuzzy_rule <- c(...)
```

**Examples**

```
# For example, the rule "IF CTemp IS warm THEN CPcomfort IS very comfortable"
# is coded as:

fuzzy_rule(0,1,0,0,0,1)

# Where, the first three values (0,1,0) correspond with the linguistic
# expressions Atemp=(cold, warm, hot) that define the room temperature (CTemp).
# The last three values (0,0,1) are related to the linguistic expressions
# Acomfort=(uncomfortable, comfortable and very comfortable) that define
# the room comfort (CPcomfort).
#
```

---

fuzzy\_rules

*Define the fuzzy rules*


---

**Description**

It is a constructor of fuzzy rules, the arguments are the different [fuzzy\\_rule](#) object.

**Usage**

```
fuzzy_rules(...)
```

**Arguments**

... one or more [fuzzy\\_rule](#) objects.

**Value**

```
fuzzy_rules <- list(...)
```

**Examples**

```
fuzzy_rules(fuzzy_rule(0,0,1, 0,0,1, 0,0,1),
            fuzzy_rule(1,0,0, 1,0,0, 1,0,0),
            fuzzy_rule(0,1,0, 0,1,0, 0,1,0))
```

---

generate_code	<i>Generate the R code</i>
---------------	----------------------------

---

**Description**

The function takes as input the path to a XML file that contains a LDCP system. Then it generates its corresponding in R code. This R code is stored in an output file. The output file path is another function parameter.

**Usage**

```
generate_code(input, output)
```

**Arguments**

input	is the XML source path file. E.g. "/folder/ldcp_system.xml".
output	is the R destination path file. E.g. "/folder/ldcp_system.R".

**Value**

If the process ends without error, the user will receive a message that indicates that the code has been generated successfully. Otherwise, the user will receive the detailed list of errors.

**Examples**

```
## Not run: generate_code('extdata/comfortableroom', 'comfortableroom')  
  
## The code has been generated successfully
```

---

glmp	<i>Define the GLMP</i>
------	------------------------

---

**Description**

Granular Linguistic Model of Phenomena (GLMP) is a network of [cp](#) and [pm](#) objects. that allows the designer to model computationally her/his own perceptions. The input data are introduced into the model through 1PMs which interpret the input data and create CPs. Then, 2PMs take several CPs as input and generate 2CPs. Of course, following the same scheme, is possible to add additional upper levels.

The `glmp` constructor receive as arguments the list of pms and the method with the computational model.

**Usage**

```
glmp(pms, method)
```

**Arguments**

`pm` is the list of `pm` objects included in the `glmp`.  
`method` is the function with the `glmp` computational model. The method must have two arguments: the list of `pm` objects defined in the `glmp` and the input data:  
`my_glmp_method <- function(pm, input)`

**Value**

The generated `glmp = list(pm, method)`

**Examples**

```
## Not run: glmp_method <- function(pm, input){

  pm$pm_depth <- pm_infer(pm$pm_depth, input[1])
  pm$pm_height <- pm_infer(pm$pm_height, input[2])
  pm$pm_width <- pm_infer(pm$pm_width, input[3])

  pm$pm_frame <- pm_infer(pm$pm_frame, list( pm$pm_depth$,
                                             pm$pm_height$,
                                             pm$pm_width$y)
  )
  pm
}

my_glmp <- glmp(list(pm_depth = pm_depth,
                    pm_height = pm_height,
                    pm_width = pm_width,
                    pm_frame = pm_frame),
               glmp_method)

## End(Not run)
```

---

infer\_rules

*Make the inference*


---

**Description**

Make an inference with the fuzzy rules.

**Usage**

```
infer_rules(rules, operator, input)
```

**Arguments**

`rules` the set of fuzzy rules.  
`operator` the `operator` object.  
`input` is the list of validity degrees related to the input `cp` objects.

**Value**

A vector that contained the result of the inference.

**Examples**

```
## In the example the input of the fuzzy rule correspond with two CPs and each CP has 3
## linguistic variables, e.g, {"bad", "good", "very good"}. The output also
## correspond with a CP with 3 linguistic variables.
```

```
infer_rules(fuzzy_rules(fuzzy_rule(0,0,1, 0,0,1, 0,0,1),
                        fuzzy_rule(1,0,0, 1,0,0, 1,0,0),
                        fuzzy_rule(0,1,0, 0,1,0, 0,1,0)),
            operator(min, max),
            list(c(0,0.5,0.5),c(0.5,0.5,0)))
## [1] 0.0 0.5 0.0
```

---

 ldcp

*Define the LDCP system*


---

**Description**

Linguistic Descriptions of Complex Phenomena (LDCP) is a technology focused on modeling complex phenomena, interpreting input data and generating automatic text reports customized to the user needs. #' The ldcp constructor receive as arguments: the `data_structure`, the `glmp` and the `report_template`.

**Usage**

```
ldcp(data, glmp, report)
```

**Arguments**

`data` is the `data_structure` object.  
`glmp` is the `glmp` object.  
`report` is the `report_template` object.

**Value**

The generated system `ldcp = list(data, glmp, report)`

**See Also**

`cp` and `pm`

**Examples**

```
## Not run: my_ldcp <- ldcp(my_data,my_glmp,my_report)
```

---

ldcp_run	<i>Execute the LDCP system</i>
----------	--------------------------------

---

### Description

Execute the `ldcp` system in order to obtain the linguistic report. This method follows these three sequential steps 1) Data acquisition, 2) Interpretation and 3) Report generation. Data acquisition process gets the input data and prepares the data structure. Then, the data are interpreted using the GLMP. The result is a set of computational perceptions (CP) that are valid to describe these data. Finally, the report generation process generates a linguistic report using the report template and the previous set of CPs.

### Usage

```
ldcp_run(ldcp, input = NULL)
```

### Arguments

<code>ldcp</code>	is the <code>ldcp</code> system.
<code>input</code>	is the system input data. May be a vector, list, or matrix with numbers. It is a new input to the <code>data_structure</code> object. By default, is NULL.

### Value

The `ldcp` object that contains the execution results.

### Examples

```
## Not run: my_ldcp <- ldcp_run(my_ldcp)
```

---

operator	<i>Define the operator</i>
----------	----------------------------

---

### Description

The operator defines the conjunction and disjunction functions used in the fuzzy rules. It takes as input parameters the function used to implement the conjunction, and the function used to implement the disjunction, e.g., "operator(min, max)", where min and max are functions defined by the R language that calculate the maximum and minimum, respectively, from a set of values received as input. Note that, we implicitly assign to the fuzzy implication operator (THEN) the function given for conjunction

### Usage

```
operator(conjunction, disjunction)
```

**Arguments**

conjunction is the method used to make the conjunction.  
 disjunction is the method used to make the disjunction.

**Value**

the operator object `my_op <- list(conjunction, disjunction)`.

**Examples**

```
operator <- operator(min, max)
```

---

pm	<i>Define the PM</i>
----	----------------------

---

**Description**

Perception Mapping (PM) is used to create and aggregate `cp` objects. Each PM receives a set of inputs (`cp` objects or numerical values) which are aggregated into a single CP.

**Usage**

```
pm(u = NULL, y, g, t = NULL)
```

**Arguments**

`u` is a vector of  $n$  input `cps`  $u = (u_1, u_2, \dots, u_n)$ . In the special case of first level perception mappings (1PM) the inputs are numerical values provided either by sensors or obtained from a database.

`y` is the output `cp`.

`g` is an aggregation function employed to calculate  $w$  from the input `cps`. For example, `g` might be implemented using a set of fuzzy rules. In the case of 1PMs, `g` is built using a set of membership functions.

`t` is a text generation algorithm which allows generating the sentences in  $A$ . In simple cases, `t` is a linguistic template, e.g., `cat("Alabama has", value, "the number of women in the last census")`.

**Value**

The generated `pm = list(u, y, g, t)`

**See Also**

[cp](#)

**Examples**

```

## Not run: cp_depth <- cp("cp_depth",c("far",
                                "bit far",
                                "good",
                                "close",
                                "very close"))

g_depth <- function(u,y){
  y$w <- degree_mf(fuzzy_partitions(triangle_mf(450,450,550),
                                     triangle_mf( 450,550,600),
                                     trapezoid_mf(550,600,800, 1000),
                                     triangle_mf( 800,1000,1300),
                                     trapezoid_mf( 1000,1300,1500,1500)),u)

  y
}

pm_depth <- pm(y=cp_depth, g=g_depth)

##### HEIGHT DEFINITION #####
cp_height <- cp("cp_height", c("high",
                                "average high",
                                "centered",
                                "average low",
                                "low"))

g_height <- function(u,y){
  y$w <- degree_mf(fuzzy_partitions(trapezoid_mf(-1000,-1000,-600,-400),
                                     triangle_mf(-600,-400,0),
                                     trapezoid_mf(-400,0,200,400),
                                     triangle_mf(200,400,600),
                                     trapezoid_mf(400,600,1000,1000)),u)

  y
}

pm_height <- pm(y=cp_height, g=g_height)

##### WIDTH DEFINITION #####
cp_width <- cp("cp_width", c("left",
                                "average left",
                                "centered",
                                "average right",
                                "right"))

g_width <- function(u,y){
  y$w <- degree_mf(fuzzy_partitions(triangle_mf(-1000,-600,-400),
                                     triangle_mf(-600,-400,0),
                                     triangle_mf(-400,0,400),
                                     triangle_mf(0,400,600),
                                     triangle_mf(400,600,1000,1000)),
                                u)

  y
}

```

```

pm_width <- pm(y=cp_width, g=g_width)

##### FRAME DEFINITION #####
cp_frame <- cp("cp_frame", c("bad",
                             "middle",
                             "good"))

g_frame <- function(u,y){

  operator <- operator(min, max)

  y$w<- infer_rules(fuzzy_rules( fuzzy_rule(0,0,1,0,0, 0,0,1,0,0, 0,0,1,0,0, 0,0,1),
                                   fuzzy_rule(1,1,1,1,1, 1,1,1,1,1, 1,1,0,1,1, 1,0,0),
                                   fuzzy_rule(1,1,1,1,1, 1,0,0,0,1, 0,0,1,0,0, 1,0,0),
                                   fuzzy_rule(1,0,0,0,1, 1,1,1,1,1, 0,0,1,0,0, 1,0,0),
                                   fuzzy_rule(0,1,0,1,0, 0,1,0,1,0, 0,0,1,0,0, 0,1,0)),
                    operator,
                    list(u[[1]]$w,u[[2]]$w,u[[3]]$w))

  y
}

t_frame <- function(y){

  templates <- c("It has been taken a bad framed photo",
                "It has been taken a middle framed photo",
                "It has been taken a good framed photo")

  return( templates[which.max(y$w)])
}

pm_frame <- pm(y=cp_frame, g=g_frame, t=t_frame)

## End(Not run)

```

---

pm\_infer

*Call the g function*


---

### Description

It call the g function in order to make the inference, i.e., map inputs u to output y.

### Usage

```
pm_infer(pm, u = NULL)
```

### Arguments

pm	is the pm object.
u	is the new pm input. By default is NULL.

**Value**

the pm obtained after calling g.

**See Also**

[cp](#)

**Examples**

```
cp_depth <- cp("cp_depth", c("far",
                             "bit far",
                             "good",
                             "close",
                             "very close"))

g_depth <- function(u,y){
  y$w <- degree_mf(fuzzy_partitions(triangle_mf(450,450,550),
                                     triangle_mf( 450,550,600),
                                     trapezoid_mf(550,600,800, 1000),
                                     triangle_mf( 800,1000,1300),
                                     trapezoid_mf( 1000,1300,1500,1500)),u)

  y
}

pm_depth <- pm(y=cp_depth, g=g_depth)
pm_depth <- pm_infer(pm_depth, 650)
```

---

pm\_multidimensional *Define the pm of a multidimensional cp*

---

**Description**

It is a set of [pms](#) that infer a multidimensional [cp](#).

**Usage**

```
pm_multidimensional(...)
```

**Arguments**

... the set of [pms](#)

**Value**

The generated pm\_multidimensional <- list(...)

---

pm_report	<i>Generate the report of y</i>
-----------	---------------------------------

---

### Description

It call the `t` function in order to generate the linguistic descriptions that better describe the output `y`.

### Usage

```
pm_report(pm)
```

### Arguments

`pm` is the pm object.

### Value

the description obtained after calling `t`.

### Examples

```
cp_frame <- cp("cp_frame", c("bad",
                             "middle",
                             "good"))

g_frame <- function(u,y){

operator <- operator(min, max)

y$w<- infer_rules(fuzzy_rules( fuzzy_rule(0,0,1,0,0, 0,0,1,0,0, 0,0,1,0,0, 0,0,1),
                                fuzzy_rule(1,1,1,1,1, 1,1,1,1,1, 1,1,0,1,1, 1,0,0),
                                fuzzy_rule(1,1,1,1,1, 1,0,0,0,1, 0,0,1,0,0, 1,0,0),
                                fuzzy_rule(1,0,0,0,1, 1,1,1,1,1, 0,0,1,0,0, 1,0,0),
                                fuzzy_rule(0,1,0,1,0, 0,1,0,1,0, 0,0,1,0,0, 0,1,0)),
operator,
list(u[[1]]$w,u[[2]]$w,u[[3]]$w))

y
}

t_frame <- function(y){

templates <- c("It has been taken a bad framed photo",
              "It has been taken a middle framed photo",
              "It has been taken a good framed photo")

return( templates[which.max(y$w)])
}

pm_frame <- pm(y=cp_frame, g=g_frame, t=t_frame)
```

```
pm_report(pm_frame)
```

---

report\_template      *Define the report template*

---

### Description

The text generation algorithm contains the programming code capable of generating the appropriate report to each specific user. Algorithms must select and order the linguistic expressions to generate the text included in the report. #' The report\_template constructor receive as arguments the list of properties and the method (programming code) capable of generating the appropriate report.

### Usage

```
report_template(properties = NULL, method, description = NULL)
```

### Arguments

properties	may be a vector, list or matrix with the user's needs, preferences and goals. By default properties = NULL.
method	is the function that generates the appropriate report. The method must have two arguments: the list of properties and the list of pms: my_report_method <- function(properties, pm){...}.
description	is the result of call the report template. By default is NULL

### Value

The generated report\_template= list(properties, method,description)

### Examples

```
properties = NULL
report_method <- function(properties,pm){
  pm_report(pm$pm_frame)
}
my_report <- report_template(properties,
                             report_method)
```

---

trapezoid_mf	<i>Define the trapezoid membership function</i>
--------------	---

---

**Description**

It is a constructor of trapezoidal shapes. They take as input the numerical values which define the anchor points in membership functions.

**Usage**

```
trapezoid_mf(a, b, c, d)
```

**Arguments**

a	the trapezoid point a.
b	the trapezoid point b.
c	the trapezoid point c.
d	the trapezoid point d.

**Value**

```
the (trapezoid_mf <- list(a,b,c,d))
```

**Examples**

```
trapezoid_mf(0, 1, 2, 3)
```

---

triangle_mf	<i>Define the triangle membership function</i>
-------------	--

---

**Description**

It is a constructor of triangular shapes. They take as input the numerical values which define the anchor points in membership functions.

**Usage**

```
triangle_mf(a, b, c)
```

**Arguments**

a	the trapezoid point a.
b	the trapezoid point b.
c	the trapezoid point c.

**Value**

```
the (triangle_mf <- list(a,b,c))
```

**Examples**

```
triangle_mf(0, 1, 2)
```

---

validate_xml	<i>Validate the XML file</i>
--------------	------------------------------

---

**Description**

The function takes as input the path to a XML file that contains a LDCP system. Then it validates the LDCP system.

**Usage**

```
validate_xml(xmlfile, schema = NULL)
```

**Arguments**

xmlfile	is the XML source path file. E.g. "/folder/ldcp_system.xml".
schema	is the ldcp schema path file. By default is "ldcpSchema.xsd".

**Value**

If the process ends without error, the user will receive the message that indicates that the XML is valid. Otherwise, the user will receive the detailed list of errors.

**Examples**

```
## Not run: validate_xml('extdata/comfortableroom.xml')  
  
## The xml is valid
```

---

`xml2rldcp`*XML to rLDCP*

---

**Description**

The function takes as input the path to a XML file that contains a LDCP system. Then it validates the LDCP system and generates its corresponding in R code. This R code is stored in an output file. The output file path is another function parameter.

**Usage**

```
xml2rldcp(input, output)
```

**Arguments**

<code>input</code>	is the XML source path file. E.g. <code>"/folder/ldcp_system.xml"</code> .
<code>output</code>	is the R destination path file. E.g. <code>"/folder/ldcp_system.R"</code> .

**Value**

If the process ends without error, the user will receive two messages: one indicates that the XML is valid and the other indicates that the code has been generated successfully. Otherwise, the user will receive the detailed list of errors.

**Examples**

```
## Not run: xml2rldcp('extdata/comfortableroom.xml', 'comfortableroom.R')  
  
## The xml is valid  
## The code has been generated successfully
```

# Index

cp, [2](#), [7-9](#), [11](#), [14](#)

data\_structure, [3](#), [9](#), [10](#)

degree\_mf, [4](#)

fuzzy\_partitions, [5](#)

fuzzy\_rule, [5](#), [6](#)

fuzzy\_rules, [6](#)

generate\_code, [7](#)

glmp, [7](#), [9](#)

infer\_rules, [8](#)

ldcp, [9](#), [10](#)

ldcp\_run, [10](#)

operator, [8](#), [10](#)

pm, [7-9](#), [11](#), [14](#)

pm\_infer, [13](#)

pm\_multidimensional, [14](#)

pm\_report, [15](#)

report\_template, [9](#), [16](#)

trapezoid\_mf, [17](#)

triangle\_mf, [17](#)

validate\_xml, [18](#)

xml2rldcp, [19](#)