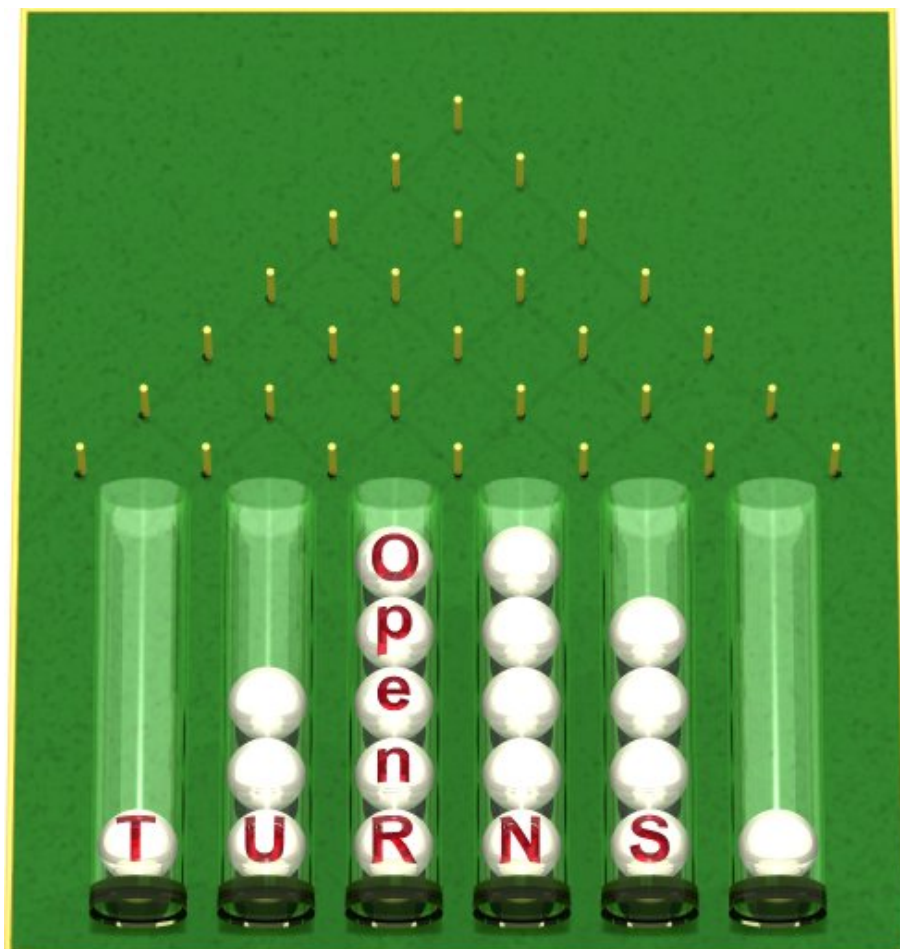


# User Manual for the Textual User Interface

Open TURNS version 0.12.1

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## 1 Base Objects

In this section a description of general objects is given. These objects are used in the different following sections.

### 1.1 CorrelationMatrix

**Usage :**

*CorrelationMatrix(dim)*

*CorrelationMatrix(dim, values)*

**Arguments :**

*dim* : an integer, the dimension of the CorrelationMatrix (square matrix with *dim* rows and *dim* colons)

*values* : a NumericalScalarCollection of dimension  $dim^2$  which contains values to put in the CorrelationMatrix, filled by rows. When these values are not specified, the CorrelationMatrix is initialised to the identity matrix.

**Value :** a CorrelationMatrix

while using the first parameters set, the correlation matrix is the identity matrix

while using second parameters set, the correlation matrix contains the specified values, filled by row

**Some methods :**

*str*

**Usage :** *str()*

**Arguments :** no argument

**Value :** a string giving the description of the (class, name, dimension, values)

*transpose*

**Usage :** *transpose()*

**Arguments :** no argument

**Value :** a CorrelationMatrix, the transposed CorrelationMatrix

*computeDeterminant*

**Usage :** *computeDeterminant()*

**Arguments :** no argument

**Value :** a real value giving the determinant of the CorrelationMatrix

*computeEigenValues*

**Usage :** *computeEigenValues()*

**Arguments :** no argument

**Value :** a NumericalPoint giving the eigen values of the CorrelationMatrix

**Links :** see docref\_B121\_ChoixLoi

## 1.2 Description

**Usage :**

*Description(dim)*

*Description(dim, name)*

**Arguments :**

*dim* : an integer, the dimension of the Description

*name* : a string to name the Description

**Value :** a Description

**Some methods :**

*[]*

**Usage :** *Description[i]*

**Arguments :** *i* : an integer, constraint :  $0 \leq i \leq dim - 1$

**Value :** a string, the description of the  $(i + 1)$ -th element of the Description

*add*

**Usage :** *add(str)*

**Arguments :** *str* : a string

**Value :** an element is added to the Description which name is *str*

*getSize*

**Usage :** *getSize()*

**Arguments :** none

**Value :** an integer, the size of the Description (*dim*)

*getName*

**Usage :** *getName()*

**Arguments :** none

**Value :** a string, the name of the Description

*setName*

**Usage :** *setName(name)*

**Arguments :** *name* : a string to name the Description

**Value :** the Description is then named *name*

### 1.3 DistributionCollection

**Usage :** *DistributionCollection(dim)*

**Arguments :** *dim* : an integer, the dimension of the collection of distributions

**Value :** a DistributionCollection, to be filled after

**Some methods :**

*[]*

**Usage :** *DistributionCollection[i]*

**Arguments :** *i* : an integer, constraint :  $0 \leq i \leq dim - 1$

**Value :** a Distribution, the  $(i + 1)$ -th Distribution of the DistributionCollection

*add*

**Usage :** *add(distribution)*

**Arguments :** *distribution*, a Distribution

**Value :** a DistributionCollection of size  $dim + 1$ , which  $(dim + 1)$ -th element is *distribution*

*at*

**Usage :** *at(i)*

**Arguments :** *i*, an integer

**Value :** a Distribution, the distribution of the  $(i + 1)$ -th component of a random vector (type `CollectionOfDistribution.at(i).str()` to have a brief description)

**Links :** see `docref_B_JoinedCDF_en`

## 1.4 HistogramPair

**Usage :** *HistogramPair*( $h, l$ )

**Arguments :**

$h$  : a real value, the height of each element of the Histogram

$l$  : a real value, the width of each element of the Histogram

**Value :** a HistogramPair

**Details :**

This object is used to build a HistogramPairCollection (hence, also used to create an Histogram)



## 1.5 HistogramPairCollection

**Usage :** *HistogramPairCollection(dim)*

**Arguments :** *dim* : an integer, the number of elements of the HistogramPairCollection

**Value :** a HistogramPairCollection, to be filled after

**Some methods :**

*[]*

**Usage :** *HistogramPairCollection[i]*

**Arguments :** *i* : an integer, must be  $< dim$

**Value :** a HistogramPair, the  $(i + 1)$ -th of the HistogramPairCollection

*add*

**Usage :** *add(HistP)*

**Arguments :** *HistP* : a HistogramPair

**Value :** an HistogramPairCollection of  $Size = dim + 1$  with instance of the  $dim + 1$ -th element of the HistogramPairCollection

*getSize*

**Usage :** *getSize()*

**Arguments :** no argument

**Value :** an integer, the size of HistogramPairCollection (returns *dim*)

*str*

**Usage :** *str()*

**Arguments :** no argument

**Value :** a string with elements of HistogramPairCollection

## 1.6 Matrix

**Usage :**

*Matrix*( $n_r, n_c$ )

*Matrix*( $n_r, n_c, values$ )

**Arguments :**

$n_r$  : an integer, the number of rows of the Matrix

$n_c$  : an integer, the number of columns of the Matrix

$values$  : a NumericalScalarCollection with  $n_r \times n_c$  elements

**Value :** a Matrix

while using the first parameters set, the matrix is filled with 0.

while using second parameters set, the Matrix contains values of the NumericalScalarCollection. The matrix is filled by row

**Some methods :**

[, ]

**Usage :** *Matrix*[ $i, j$ ]

**Arguments :**

$i$  : an integer, constraint :  $0 \leq i \leq n_r - 1$

$j$  : an integer, constraint :  $0 \leq j \leq n_c - 1$

**Value :** a real value, the  $(i, j)$  element of the Matrix

*getNbColumns*

**Usage :** *getNbColumns*()

**Arguments :** none

**Value :** an integer : the number of column  $n_c$

*getNbRows*

**Usage :** *getNbRows*()

**Arguments :** none

**Value :** an integer : the number of row  $n_r$

*transpose*

**Usage :** *transpose*()

**Arguments :** none

**Value :** the transposed Matrix

*getName*

**Usage :** *getname*()

**Arguments :** none

**Value :** a string, the name of the Matrix

*setName*

**Usage :** *setname(name)*

**Arguments :** name : a string

**Value :** the Matrix is named *name*

*solveLinearSystem*

**Usage :** *solveLinearSystem(y)*

**Arguments :** *y* a NumericalPoint of dimension  $n_r$  (the number of rows of the Matrix)

**Value :** NumericalPoint, *x*, such that

$\text{Matrix} * \mathbf{x} = \mathbf{y}$

## 1.7 NumericalMathFunction

**Usage :**

*NumericalMathFunction(file)*  
*NumericalMathFunction(input, output, formula)*  
*NumericalMathFunction(f, g)*

**Arguments :**

*input* : a Description which describes the input of the NumericalMathFuction  
*output* : a Description which describes the output of the NumericalMathFuction  
*formula* : a Description, the analytical formula of the NumericalMathFuction  
*file* : a string to name the XML file (without the extension ".xml") which contains the implementation of the considered function  
*f, g* : two NumericalMathFunction in norder to create the composition function  $f \circ g$

**Value :** NumericalMathFunction

**Some methods :**

()

**Usage :**

*NumericalMathFunction(x)*  
*NumericalMathFunction(sample)*

**Arguments :**

*x* : a NumericalPoint  
*sample* : a NumericalSample

**Value :**

while using the first usage, a NumericalPoint, the NumericalMathFunction value at point *x*  
 while using the second usage, a NumericalSample, the NumericalMathFunction value on the sample *sample*

*getDescription*

**Usage :** *getDescription()*

**Arguments :** none

**Value :** a Description which describes the inputs and the ouputs of the NumericalMathFunction  
 (use `print NumericalMathFuction.getDescription()` command to visualize it)

*getEvaluationCallsNumber*

**Usage :** *getEvaluationCallsNumber()*

**Arguments :** none

**Value :** an integer that counts the number of times the NumericalMathFunction has been called since the beginning of the python session

*getGradientCallsNumber*

**Usage :** *getGradientCallsNumber()*

**Arguments :** none

**Value :** an integer that counts the number of times the gradient of the NumericalMathFunction has been called since the beginning of the python session. Note that if the gradient is implemented by a finite difference method, the gradient calls numbers is equal to 0 and the different calls are comptabilised in the evaluation calls number

*getHessianCallsNumber*

**Usage :** *getHessianCallsNumber()*

**Arguments :** none

**Value :** an integer that counts the number of times the gradient of the NumericalMathFunction has been called since the beginning of the python session. Note that if the hessian is implemented by a finite difference method, the hessian calls numbers is equal to 0 and the different calls are comptabilised in the evaluation calls number

*getInputDescription*

**Usage :** *getInputDescription()*

**Arguments :** none

**Value :** a Description which describes the inputs of the NumericalMathFunction

*getInputNumericalPointDimension* or *getInputDimension*

**Usage :** *getInputNumericalPointDimension(),getInputDimension()*

**Arguments :** none

**Value :** an integer, the dimension of the input space

*getMarginal*

**Usage :** *getMarginal(i)*

**Arguments :** *i* : an integer corresponding to the marginal (Care : the numerotation begins at 0 )

**Value :** a NumericalMathFunction, noted  $f_i$  if  $f : \mathcal{R}^n \longrightarrow \mathcal{R}^p$ , with  $f = (f_0, \dots, f_p)$ .

*getOutputDescription*

**Usage :** *getOutputDescription()*

**Arguments :** none

**Value :** a Description which describes the outputs of the NumericalMathFunction object

*getOutputNumericalPointDimension* or *getOutputDimension*

**Usage :** *getOutputNumericalPointDimension(),getOutputDimension()*

**Arguments :** none

**Value :** an integer, the dimension of the output space

*getParameters*

**Usage :** *getParameters()*

**Arguments :** none

**Value :** a NumericalPoint, the NumericalPoint corresponding to parameters of the NumericalMathFunction

#### *GetValidOperators*

**Usage :** *GetValidOperators()*

**Arguments :** none

**Value :** a Description, containing the list of the operators we can use within Open TURNS

#### *GetValidFunctions*

**Usage :** *GetValidFunctions()*

**Arguments :** none

**Value :** a Description, containing the list of the functions we can use within Open TURNS

#### *GetValidConstants*

**Usage :** *GetValidConstants()*

**Arguments :** none

**Value :** a Description, containing the list of the constants we can use within Open TURNS

#### *gradient*

**Usage :** *gradient(x)*

**Arguments :** *x* : a NumericalPoint (which has the same dimension as the inputs)

**Value :** a Matrix, the gradient (with respect to the inputs) of the NumericalMathFunction

#### *hessian*

**Usage :** *hessian(x)*

**Arguments :** *x* : a NumericalPoint (which has the same dimension as the inputs)

**Value :** a SymmetricTensor, the hessian (with respect to the inputs) of the NumericalMathFunction

#### *setName*

**Usage :** *setName(name)*

**Arguments :** *name* : a string (between quotations marks)

**Value :** it gives a name to the NumericalMathFunction

Here is the list of constants proposed by Open TURNS :

- *\_e* : Euler's constant (2.71828...),
- *\_pi* : Pi constant (3.14159...)

Here is the list of functions proposed by Open TURNS :

- *sin(arg)* : sine function,
- *cos(arg)* : cosine function,

- $\tan(arg)$  : tangent function,
- $\sin(arg)$  : inverse sine function,
- $\cos(arg)$  : inverse cosine function,
- $\tan^{-1}(arg)$  : inverse tangent function,
- $\sinh(arg)$  : hyperbolic sine function,
- $\cosh(arg)$  : hyperbolic cosine function,
- $\tanh(arg)$  : hyperbolic tangens function,
- $\sinh^{-1}(arg)$  : inverse hyperbolic sine function,
- $\cosh^{-1}(arg)$  : inverse hyperbolic cosine function,
- $\tanh^{-1}(arg)$  : inverse hyperbolic tangent function,
- $\log_2(arg)$  : logarithm in base 2,  $\log_{10}(arg)$  : logarithm in base 10,  $\log(arg)$  : logarithm in base e (2.71828...),  $\ln(arg)$  : alias for log function,
- $\ln\Gamma(arg)$  : log of the gamma function,
- $\Gamma(arg)$  : gamma function,
- $\exp(arg)$  : exponential function,
- $\operatorname{erf}(arg)$  : error function,
- $\operatorname{erfc}(arg)$  : complementary error function,
- $\sqrt{arg}$  : square root function,
- $\sqrt[3]{arg}$  : cubic root function,
- $\text{besselJ0}(arg)$  : 1st kind Bessel function with parameter 0,
- $\text{besselJ1}(arg)$  : 1st kind Bessel function with parameter 1,
- $\text{besselY0}(arg)$  : 2nd kind Bessel function with parameter 0,
- $\text{besselY1}(arg)$  : 2nd kind Bessel function with parameter 1,
- $\text{sign}(arg)$  : sign function -1 if  $x < 0$ ; 1 if  $x > 0$ ,
- $\text{rint}(arg)$  : round to nearest integer function,
- $\text{abs}(arg)$  : absolute value function,
- $\text{if}(arg1, arg2, arg3)$  : if  $arg1$  then  $arg2$  else  $arg3$ ,
- $\min(arg1, \dots, argn)$  : min of all arguments,
- $\max(arg1, \dots, argn)$  : max of all arguments,
- $\text{sum}(arg1, \dots, argn)$  : sum of all arguments,

- $avg(arg1, ..., argn)$  : mean value of all arguments .

Here is the list of operators proposed by Open TURNS :

- $=$  : assignement, can only be applied to variable names (priority -1),
- $and$  : logical and (priority 1),
- $or$  : logical or (priority 1),
- $xor$  : logical xor (priority 1),
- $\leq$  : less or equal (priority 2),
- $\geq$  : greater or equal (priority 2),
- $\neq$  : not equal (priority 2),
- $==$  : equal (priority 2),
- $>$  : greater than (priority 2),
- $<$  : less than (priority 2),
- $+$  : addition (priority 3),
- $-$  : subtraction (priority 3),
- $*$  : multiplication (priority 4),
- $/$  : division (priority 4),
- $\neg$  : logical negation (priority 4),
- $not$  : alias for  $\neg$  (priority 4),
- $-$  : sign change (priority 4),
- $^$  : raise x to the power of y (priority 5).

## 1.8 NumericalPoint

**Usage :** *NumericalPoint(dim, value)*

**Arguments :**

*dim* : an integer, the dimension of the NumericalPoint

*value* : a real value, the value of each component of the NumericalPoint

**Some methods :**

$[]$

**Usage :** *NumericalPoint[i]*

**Arguments :**

*i* : an integer, constraint :  $0 \leq i \leq dim - 1$



**Value :** a real value, the value of the  $(i + 1)$ -th element of the NumericalPoint

*getDimension*

**Usage :** *getDimension()*

**Arguments :** none

**Value :** an integer, the value of the dimension of the NumericalPoint (it returns dim)

*norm*

**Usage :** *norm()*

**Arguments :** none

**Value :** a real value, the euclidian norm of the NumericalPoint

*norm2*

**Usage :** *norm2()*

**Arguments :** none

**Value :** a real value, the square of the euclidian norm of the NumericalPoint

*str*

**Usage :** *str()*

**Arguments :** none

**Value :** a string describing the NumericalPoint

*dot*

**Usage :** *dot(x, y)*

**Arguments :**  $x, y$  : NumericalPoint

**Value :** a real value, the dot product (also known as the scalar product) of  $x$  and  $y$

*getName*

**Usage :** *getName()*

**Arguments :** none

**Value :** a string giving the name of the NumericalPoint

*setName*

**Usage :** *setName(name)*

**Arguments :**  $name$  : a string

**Value :** no value, it gives a name for the considered NumericalPoint

## 1.9 NumericalPointCollection

**Usage :** *NumericalPointCollection(dim)*

**Arguments :** *dim* : an integer, the number of elements of the NumericalPointCollection

**Value :** a NumericalPointCollection, filled by default with 0.0

**Some methods :**

*[]*

**Usage :** *NumericalPointCollection[i]*

**Arguments :** *i* : an integer, constraint :  $0 \leq i \leq dim - 1$

**Value :** a NumericalPoint, the  $(i + 1)$ -th element of the NumericalPointCollection

*add*

**Usage :** *add(NumericalPoint2)*

*NumericalPoint* : a NumericalPoint

**Value :** The NumericalPointCollection of size  $dim + 1$ . The  $dim + 1$  element of this object is then equal to *NumericalPoint2*

*getSize*

**Usage :** *getSize()*

**Arguments :** none

**Value :** an integer : the size of the NumericalPointCollection

### 1.10 NumericalScalarCollection

**Usage :** *NumericalScalarCollection(dim)*

**Arguments :** *dim* : an integer, the number of elements of the NumericalScalarCollection

**Value :** a NumericalScalarCollection, filled by default with 0

**Some methods :**

*[]*

**Usage :** *NumericalScalarCollection[i]*

**Arguments :**

*i* : an integer, constraint :  $0 \leq i \leq dim - 1$

**Value :** a real value, the  $(i + 1)$ -th element of the NumericalScalarCollection

*add*

**Usage :** *add(val)*

*val* : a real value

**Value :** The NumericalScalarCollection of size  $dim + 1$ . The  $dim + 1$  element of this object is then equal to *val*

*getSize*

**Usage :** *getSize()*

**Arguments :** none

**Value :** an integer : the size of the NumericalScalarCollection

## 1.11 SquareMatrix

**Usage :**

*SquareMatrix(dim)*  
*SquareMatrix(dim, values)*

**Arguments :**

*dim* : an integer, the dimension of the SquareMatrix (square matrix with *dim* rows and *dim* colons)  
*values* : a NumericalScalarCollection of dimension  $dim^2$

**Value :** SquareMatrix

while using the first parameters set, the SquareMatrix is filled with 0.

while using the second parameters set, the SquareMatrix contains values of the NumericalScalarCollection. SquareMatrix is filled by rows.

**Some methods :**

*computeDeterminant*

**Usage :** *computeDeterminant()*

**Arguments :** none

**Value :** a real value giving the determinant of the SquareMatrix

*computeEigenValues*

**Usage :** *computeEigenValues()*

**Arguments :** none

**Value :** a NumericalPoint giving the eigen values of the SquareMatrix

*getDimension*

**Usage :** *getDimension()*

**Arguments :** none

**Value :** an integer, the dimension of the SquareMatrix (it returns *dim*)

*solveLinearSystem*

**Usage :** *solveLinearSystem(y)*

**Arguments :** *y* : a NumericalPoint of dimension  $n_r$  (the number of row of the SquareMatrix)

**Value :** NumericalPoint, this NumericalPoint, *x*, is such that

SquareMatrix \* *x* = *y*

*transpose*

**Usage :** *transpose()*

**Arguments :** none

**Value :** the transposed SquareMatrix

## 1.12 Tensor

**Usage :**

*Tensor*( $n_r, n_c, n_s$ )

*Matrix*( $n_r, n_c, n_s, values$ )

**Arguments :**

$n_r$  : an integer, the number of row of the Tensor

$n_c$  : an integer, the number of row of the Tensor

$n_s$  : an integer, the number of sheet of the Tensor

$values$  : NumericalScalarCollection with  $n_r \times n_c \times n_s$  elements

**Value :** Tensor

while using the first parameters set, the matrix is filled with 0.

while using the second parameters set, the Matrix contains values of the NumericalScalarCollection. The tensor is filled by row.

**Some methods :**

[ , , ]

**Usage :** *Tensor*[ $i, j, k$ ]

**Arguments :**

$i$  : an integer, constraint :  $0 \leq i \leq n_r - 1$

$j$  : an integer, constraint :  $0 \leq j \leq n_c - 1$

$k$  : an integer, constraint :  $0 \leq k \leq n_s - 1$

**Value :** a real value, the  $(i, j, k)$  element of the Tensor

*getNbColumns*

**Usage :** *getNbColumns*()

**Arguments :** none

**Value :** an integer : the number of column  $n_c$

*getNbRows*

**Usage :** *getNbRows*()

**Arguments :** none

**Value :** an integer : the number of row  $n_r$

*getNbSheets*

**Usage :** *getNbSheets*()

**Arguments :** none

**Value :** an integer : the number of sheet  $n_s$

*getName*

**Usage :** *getName()*

**Arguments :** none

**Value :** a string, the name of the Tensor

*setName*

**Usage :** *setName(name)*

**Arguments :** name : a string

**Value :** the Tensor is named *name*

### 1.13 UserDefinedPair

**Usage :** *UserDefinedPair*( $x, p$ )

**Arguments :**

$x$  : a NumericalPoint,

$p$  : a real value, constraint  $0 \leq p \leq 1$  (the probabily associated to the point  $x$ )

**Value :** a UserDefinedPair

**Some methods :**

*getX()*

**Usage :** *getX()*

**Arguments :** no argument

**Value :** a NumericalPoint, the point of the UserDefinedPair

*getP()*

**Usage :** *getP()*

**Arguments :** no argument

**Value :** a NumericalScalar, the scalar of the UserDefinedPair

Each get method is associated to a set method.

### 1.14 UserDefinedPairCollection

**Usage :** *UserDefinedPairCollection(dim)*

**Arguments :** *dim* : an integer, the number of elements of the UserDefinedPairCollection

**Value :** an UserDefinedPairCollection, to be filled after

**Some methods :**

*[]*

**Usage :** *UserDefinedPairCollection[i]*

**Arguments :** *i* : an integer, the *i*th element of UserDefinedPairCollection

**Value :** a UserDefinedPair, the  $(i + 1)$ -th element of UserDefinedPairCollection

*add*

**Usage :** *add(UseDefP)*

**Arguments :** *UseDefP* : an UserDefinedPair

**Value :** a UserDefinedPairCollection of size  $dim + 1$  with instance of the  $(dim + 1)$  element of the UserDefinedPairCollection

*getSize*

**Usage :** *getSize()*

**Arguments :** no argument

**Value :** an integer, the size of UserDefinedPairCollection (returns *dim*)

*str*

**Usage :** *str()*

**Arguments :** no argument

**Value :** a string with elements of UserDefinedPairCollection



## 2 Distributions

In this section, a description about the use of Distribution object is given.

Be aware of the fact that for some uses in the TUI, it is necessary to explicitly cast a given distribution into the general Distribution class.

### 2.1 Distribution

**Usage :** *Distribution(dist, name)*

**Arguments :**

*dist* : a DistributionImplementation which is Beta, Exponential, Gamma, Geometric, Gumbel, Histogram, Logistic, LogNormal, MultiNomial, Normal, Non Central Student, Poisson, Student, Triangular, TruncatedNormal, Weibull, UserDefined,

*name* : a string to name the distribution

**Value :** a Distribution

**Some methods :**

*computeCDF*

**Usage :**

*computeCDF(value)*

*computeCDF(x)*

*computeCDF(sample)*

**Arguments :**

*x* : a NumericalScalar

*x* : a NumericalPoint

*sample* : a NumericalSample

**Value :**

while using the first usage, a NumericalScalar, the CDF (Cumulative Distribution Function) of dimension 1 value of the considered distribution at *value*

while using the second usage, a NumericalPoint, the CDF (Cumulative Distribution Function) value of the considered distribution at the vector *x*

while using the third usage, a NumericalSample, the CDF (Cumulative Distribution Function) values of the considered distribution at *sample*

*computeCDFGradient*

**Usage :** *computeCDFGradient(x)*

**Arguments :** *x* : a NumericalPoint

**Value :** a NumericalPoint object, the gradient of the distribution CDF, with respect to the parameters of the distribution, evaluated at point *x*

*computeDDF*

**Usage :**

*computeDDF(x)*  
*computeDDF(sample)*

**Arguments :**

*x* : a NumericalPoint  
*sample* : a NumericalSample

**Value :**

while using the first usage, a NumericalPoint value, the gradient of the PDF (Probability Distribution Function) of the considered distribution at *x* (DDF = Derivative Density Function)  
 while using the second usage, a NumericalSample, the gradient of the PDF (Probability Distribution Function) of the considered distribution at *x* (DDF = Derivative Density Function)

*computePDF*

**Usage :**

*computePDF(value)*  
*computePDF(x)*  
*computePDF(sample)*

**Arguments :**

*x* : a NumericalPoint  
*sample* : a NumericalSample

**Value :**

while using the first usage, a NumericalScalar, the PDF (Cumulative Distribution Function) of dimension 1 value of the considered distribution at *value*  
 while using the second usage, a NumericalPoint, the PDF (Cumulative Distribution Function) value of the considered distribution at the vector *x*  
 while using the third usage, a NumericalSample, the PDF (Cumulative Distribution Function) values of the considered distribution at *sample*

*computePDFGradient*

**Usage :** *computePDFGradient(x)*

**Arguments :** *x* : a NumericalPoint

**Value :** a NumericalPoint object, the gradient of the distribution PDF, with respect to the parameters of the distribution, evaluated at point *x*

*computeQuantile*

**Usage :** *computeQuantile(x)*

**Arguments :** *x* : a real scalar  $0 \leq x \leq 1$

**Value :** a NumericalPoint, the value of the *x*– quantile

*drawCDF*

**Usage :**

*drawCDF()*  
*drawCDF(min, max)*  
*drawCDF(min, max, pointNumber)*

*drawCDF(vectMin, vectMax)*

*drawCDF(vectMin, vectMax, vectPointNumber)*

**Arguments :**

*min* and *max* : real values with  $min < max$ , the range for the CDF curve of a distribution of dimension 1

*pointNumber* : an integer, the number of points to draw the CDF iso-curves of a distribution of dimension 1

*vectMin* and *vectMax* : two NumericalPoint of dimension 2, respectively the left-bottom and ritgh-up corners of the square for the CDF iso-curves of a distribution of dimension 2

*vectPointNumber* : a NumericalPoint of dimension 2, the the number of points to draw the iso-curves of a distribution of dimension 2 on each direction

**Value :** a Graph, containing the elements of the curve or iso-curves of the CDF, depending on the dimension of the distribution (1 or 2)

*drawPDF*

**Usage :**

*drawPDF()*

*drawPDF(min, max)*

*drawPDF(min, max, pointNumber)*

*drawPDF(vectMin, vectMax)*

*drawPDF(vectMin, vectMax, vectPointNumber)*

**Arguments :**

*min* and *max* : real values with  $min < max$ , the range for the PDF curve of a distribution of dimension 1

*pointNumber* : an integer, the number of points to draw the PDF iso-curves of a distribution of dimension 1

*vectMin* and *vectMax* : two NumericalPoint of dimension 2, respectively the left-bottom and ritgh-up corners of the square for the PDF iso-curves of a distribution of dimension 2

*vectPointNumber* : a NumericalPoint of dimension 2, the number of points to draw the iso-curves of a distribution of dimension 2 on each direction

**Value :** a Graph, containing the elements of the curve or iso-curves of the PDF, depending on the dimension of the distribution (1 or 2)

*drawMarginal1DCDF*

**Usage :**

*drawMarginal1DCDF(i, min, max, pointNumber)*

**Arguments :**

*i* : an integer, the marginal we want to draw (Care : numerotation begins at 0)

*min* and *max* : real values with  $min < max$ , the range for the CDF curve of a distribution of dimension  $>1$

*pointNumber* : an integer, the number of points to draw the CDF iso-curves of a distribution of dimension  $>1$

**Value :** a Graph, containing the elements of the curve of the CDF of the marginal *i* of the distribution of dimension  $>1$

*drawMarginal1DPDF***Usage :**

*drawMarginal1DPDF(i, min, max, pointNumber)*

**Arguments :**

*i* : an integer, the marginal we want to draw (Care : numerotation begins at 0)

*min* and *max* : real values with  $min < max$ , the range for the PDF curve of a distribution of dimension  $>1$

*pointNumber* : an integer, the number of points to draw the PDF iso-curves of a distribution of dimension  $>1$

**Value :** a Graph, containing the elements of the curve of the PDF of the marginal *i* of the distribution of dimension  $>1$

*drawMarginal2DCDF***Usage :**

*drawMarginal2DCDF(i, j, vectMin, vectMax, vectPointNumber)*

**Arguments :**

*i* and *j* : two integer, the marginal we want to draw (Care : numerotation begins at 0)

*vectMin* and *vectMax* : two NumericalPoint of dimension  $n>2$ , respectively the left-bottom and ritgh-up corners of the square for the PDF iso-curves of a distribution of dimension *n*

*vectPointNumber* : a NumericalPoint of dimension  $n>2$ , the number of points to draw the iso-curves of a distribution of dimension *n* on each direction

**Value :** a Graph, containing the elements of the iso-curve of the CDF of the marginals (*i,j*) of distribution of dimension  $n>2$

*drawMarginal2DPDF***Usage :**

*drawMarginal2DPDF(i, j, vectMin, vectMax, vectPointNumber)*

**Arguments :**

*i* and *j* : two integer, the marginal we want to draw (Care : numerotation begins at 0)

*vectMin* and *vectMax* : two NumericalPoint of dimension  $n>2$ , respectively the left-bottom and ritgh-up corners of the square for the PDF iso-curves of a distribution of dimension *n*

*vectPointNumber* : a NumericalPoint of dimension  $n>2$ , the number of points to draw the iso-curves of a distribution of dimension *n* on each direction

**Value :** a Graph, containing the elements of the iso-curve of the PDF of the marginals (*i,j*) of distribution of dimension  $n>2$

*getCopula*

**Usage :** *getCopula()*

**Arguments :** no argument

**Value :** a Copula, the copula of the considered distribution which must be of type ComposedDistribution

*getCovariance*

**Usage :** *getCovariance()*

**Arguments :** no argument

**Value :** a CovarianceMatrix of the considered distribution (if the distribution is unidimensional, it is the variance)

#### *getMarginal*

**Usage :**

*getMarginal(i)*

*getMarginal(indices)*

**Arguments :**

*i* : an integer (*i* is lower or equal to the dimension of the considered distribution), with  $0 \leq i$

*indices* : a Indices, which regroup all the indices considered

**Value :** a Distribution, the distribution of an extracted vector of the initial distribution

#### *getKurtosis*

**Usage :** *getKurtosis()*

**Arguments :** no argument

**Value :** a NumericalPoint, the value the kurtosis of each 1D marginal of the distribution

#### *getMean*

**Usage :** *getMean()*

**Arguments :** no argument

**Value :** a NumericalPoint, the value of the considered distribution mean

#### *getNumericalSample*

**Usage :** *getNumericalSample(n)*

**Arguments :** *n* : integer, the size of the sample

**Value :** a NumericalSample representing *n* realizations of the random variable with the considered distribution

#### *getParametersCollection*

**Usage :** *getParametersCollection()*

**Arguments :** one

**Value :** a NumericalPointCollection, the list of the parameters of the distribution

#### *getRealization*

**Usage :** *getRealization()*

**Arguments :** no argument

**Value :** a NumericalPoint, one realization of random variable with the considered distribution

#### *getRoughness*

**Usage :** *getRoughness()*

**Arguments :** no argument

**Value :** a NumericalScalar, the value  $roughness(\underline{X}) = ||p||_{\mathcal{L}^2} = \sqrt{\int_{\underline{x}} p^2(\underline{x}) d\underline{x}}$

*getSkewness*

**Usage :** *getSkewness()*

**Arguments :** no argument

**Value :** a NumericalPoint, the value the standard deviation of each 1D marginal of the distribution

*getStandardDeviation*

**Usage :** *getStandardDeviation()*

**Arguments :** no argument

**Value :** a NumericalPoint, the value the standard deviation of each 1D marginal of the distribution

*getWeight*

**Usage :** *getWeight()*

**Arguments :** no argument

**Value :** a NumericalScalar between 0 and 1, the weight of the considered distribution if used in a Mixture

*hasEllipticalCopula*

**Usage :** *hasEllipticalCopula()*

**Arguments :** no argument

**Value :** a boolean, it says if the considered distribution is elliptical

*hasIndependentCopula*

**Usage :** *hasIndependentCopula()*

**Arguments :** no argument

**Value :** a boolean which indicates wether the considered distribution is independent

*isElliptical*

**Usage :** *isElliptical()*

**Arguments :** no argument

**Value :** a boolean which indicates wether the considered distribution has an elliptical distribution

*str*

**Usage :** *str()*

**Arguments :** no argument

**Value :** a string describing the object

## 2.2 Usual Distributions

### 2.2.1 Beta

This class inherits from the Distribution class.

#### Usage :

Main parameters set :  $Beta(r, t, a, b)$

Second parameters set :  $Beta(\mu, \sigma, a, b, Beta.MUSIGMA)$

Default construction :  $Beta()$

#### Arguments :

$r$  : real value, first shape parameter, constraint :  $r > 0$

$t$  : real value, second shape parameter, constraint :  $t > r$

$a$  : real value, lower bound

$b$  : real value, upper bound, constraint :  $b > a$

$\mu$  : real value, mean value

$\sigma$  : real value, standard deviation, constraint :  $\sigma > 0$

**Value :** a Beta. In the default construction, we use the  $Beta(r, t, a, b) = Beta(2, 4, -1, 1)$  definition.

#### Some methods :

*getA*

**Usage :** *getA()*

**Arguments :** none

**Value :** a real value, the  $a$  parameter of the Beta distribution

*getB*

**Usage :** *getB()*

**Arguments :** none

**Value :** a real value, the  $b$  parameter of the Beta distribution

*getMu*

**Usage :** *getMu()*

**Arguments :** none

**Value :** a real value, the  $\mu$  parameter of the distribution

*getSigma*

**Usage :** *getSigma()*

**Arguments :** none

**Value :** a real value, the  $\sigma$  parameter of the distribution

*getR***Usage :** *getR()***Arguments :** none**Value :** a real value, the  $r$  parameter of the distribution*getT***Usage :** *getT()***Arguments :** none**Value :** a real value, the  $t$  parameter of the distribution**Details :**

density probability function :

$$f(x) = \frac{(x-a)^{(r-1)}(b-x)^{(t-r-1)}}{(b-a)^{(t-1)}B(r, t-r)} \mathbf{1}_{[a,b]}(x)$$

relation between parameters set :

$$\begin{aligned} \mu &= a + \frac{(b-a)r}{t} \\ \sigma &= \frac{(b-a)}{t} \sqrt{\frac{r(t-r)}{(t+1)}} \end{aligned}$$

**Links :** see `docref_B121_ChoixLoi`Each *getMethod* is associated to a *setMethod*.



### 2.2.2 Exponential

This class inherits from the `Distribution` class.

**Usage :** Main parameters set :  $Exponential(\lambda, \gamma)$

Default construction :  $Exponential()$

**Arguments :**

$\lambda$  : real value, constraint :  $\lambda > 0$

$\gamma$  : real value

**Value :** an Exponential. In the default construction, we use the  $Exponential(lambda, gamma) = Exponential(1.0, 0.0)$  definition.

**Some methods :**

*getGamma*

**Usage :** *getGamma()*

**Arguments :** none

**Value :** a real value, the  $\gamma$  parameter of the distribution

*getLambda*

**Usage :** *getLambda()*

**Arguments :** none

**Value :** a real value, the  $\lambda$  parameter of the distribution

**Details :**

density probability function :

$$f(x) = \lambda e^{-\lambda(x-\gamma)} \mathbf{1}_{[\gamma, +\infty[}(x)$$

relation between parameters set :

$$\begin{aligned} \mu &= \frac{1}{\lambda} \quad \text{where} \quad \mu = \mathbb{E}[X] \\ \sigma &= \frac{1}{\lambda} \quad \text{where} \quad \sigma = \sqrt{\text{Var}[X]} \end{aligned}$$

**Links :** voir docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.3 Gamma

This class inherits from the Distribution class.

#### Usage :

Main parameters set :  $\text{Gamma}(k, \lambda, \gamma)$

Second parameters set :  $\text{Gamma}(\mu, \sigma, \gamma, \text{Gamma.MUSIGMA})$

Default construction :  $\text{Gamma}()$

#### Arguments :

$k$  : integer value constraint :  $k > 0$

$\lambda$  : real value, constraint :  $\lambda > 0$

$\gamma$  : real value,

$\mu$  : real value, mean value

$\sigma$  : real value, standard deviation, constraint :  $\sigma > 0$

**Value :** a Gamma. In the default construction, we use the  $\text{Gamma}(k, \text{lambda}, \text{gamma}) = \text{Gamma}(1.0, 1.0, 0.0)$  definition.

#### Some methods :

*getGamma*

**Usage :** *getGamma()*

**Arguments :** none

**Value :** a real value, the  $\gamma$  parameter of the distribution

*getK*

**Usage :** *getK()*

**Arguments :** none

**Value :** a real value, the  $k$  parameter of the distribution

*getLambda*

**Usage :** *getLambda()*

**Arguments :** none

**Value :** a real value, the  $\lambda$  parameter of the distribution

#### Details :

density probability function :

$$f(x) = \frac{\lambda}{\Gamma(k)} (\lambda(x - \gamma))^{(k-1)} e^{-\lambda(x-\gamma)} \mathbf{1}_{[\gamma, +\infty[}(x)$$

relation between parameters set :

$$\begin{aligned}\mu &= \frac{k}{\lambda} + \gamma \quad \text{where} \quad \mu = \text{E}[X] \\ \sigma &= \frac{\sqrt{k}}{\lambda} \quad \text{where} \quad \sigma = \sqrt{\text{Var}[X]}\end{aligned}$$

**Links :** voir docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.4 Geometric

This class inherits from the Distribution class.

**Usage :** Main parameters set : *Geometric*( $p$ )

**Arguments :**  $p$  : a real value, constraint :  $0 < p < 1$

**Value :** Geometric

**Some methods :**

*getP*

**Usage :** *getP*()

**Arguments :** none

**Value :** a real positive value  $< 1$ , the  $p$  parameter of the distribution

**Details :**

probability distribution:

$$\mathbb{P}(k) = (1 - p)^{k-1}p, k \in \mathbb{N}^*$$

relation between parameters set :

$$\begin{aligned} \mu &= \frac{1}{p} \quad \text{where} \quad \mu = \mathbb{E}[X] \\ \sigma &= \sqrt{\frac{1-p}{p^2}} \quad \text{where} \quad \sigma = \sqrt{\text{Var}[X]} \end{aligned}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.5 Gumbel

This class inherits from the Distribution class.

#### Usage :

Main parameters set :  $Gumbel(\alpha, \beta)$

Second parameters set :  $Gumbel(\mu, \sigma, 1)$

Default construction :  $Gumbel()$

#### Arguments :

$\alpha$  : a real value, the scale parameter (the inverse), constraint :  $\alpha > 0$

$\beta$  : a real value, location parameter

$\mu$  : a real value, the mean value

$\sigma$  : a real value, standard deviation, constraint :  $\sigma > 0$

**Value :** a Gumbel. In the default construction, we use the  $Gumbel(alpha, beta) = Gumbel(1.0, 1.0)$  definition.

#### Some methods :

##### *getAlpha*

**Usage :** *getAlpha()*

**Arguments :** none

**Value :** a real value, the  $\alpha$  of the considered distribution

##### *getBeta*

**Usage :** *getBeta()*

**Arguments :** none

**Value :** a real value, the  $\beta$  of the considered distribution

##### *getMu*

**Usage :** *getMu()*

**Arguments :** none

**Value :** a real value, the  $\mu$  parameter of the distribution

##### *getSigma*

**Usage :** *getSigma()*

**Arguments :** none

**Value :** a real value, the  $\sigma$  parameter of the distribution

#### Details :

density probability function :

$$f(x) = \alpha e^{-\alpha(x-\beta)} - e^{-\alpha(x-\beta)}$$

relation between parameters set :

$$\begin{aligned}\mu &= \beta + \frac{c}{\alpha} && \text{where } c \text{ is the Euler-Mascheroni constant} && (c \approx 0.5772156649) \\ \sigma &= \frac{1}{\sqrt{6}} \frac{\pi}{\alpha}\end{aligned}$$

$$\text{where} \quad \mu = \text{E}[X] \quad \sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.6 Histogram

This class inherits from the Distribution class.

#### Usage :

Main parameters set : *Histogram(first, Coll)*

#### Arguments :

*first* : a real value, the upper bound of the distribution

*Coll* : an HistogramPairCollection, the collection of  $(h_i, l_i)$  where  $h_i$  is the height and  $l_i$  the width of each barplot of the Histogram

**Value** : an Histogram with normalized heights

#### Some methods :

*getFirst*

**Usage** : *getFirst()*

**Arguments** : none

**Value** : a real value, the *first* parameter of the considered distribution

*getPairCollection*

**Usage** : *getPairCollection()*

**Arguments** : none

**Value** : a HistogramPairCollection, the *Coll* parameter of the considered distribution

#### Details :

density probability function :

$$f(x) = \sum_{i=1}^n H_i \mathbf{1}_{[x_i, x_{i+1}]}(x)$$

where

$H_i = h_i/S$  is the normalized heights, with  $S = \sum_{i=1}^n h_i l_i$  being the initial surface of the histogram.

$l_i = x_{i+1} - x_i$ ,  $1 \leq i \leq n$

$n$  is the size of the HistogramPairCollection

Each *getMethod* is associated to a *setMethod*.

### 2.2.7 Logistic

This class inherits from the Distribution class.

#### Usage :

Main parameters set :  $Logistic(\alpha, \beta)$

Default construction :  $Logistic()$

#### Arguments :

$\alpha$  : a real value, mean value

$\beta$  : a real value, scale parameter, constraint :  $\beta \geq 0$

**Value :** Logistic. In the default construction, we use the  $Logistic(alpha, beta) = Logistic(0.0, 1.0)$  definition.

#### Some methods :

*getAlpha*

**Usage :** *getAlpha()*

**Arguments :** none

**Value :** a real value, the  $\alpha$  parameter of the considered distribution

*getBeta*

**Usage :** *getBeta()*

**Arguments :** none

**Value :** a real value, the  $\beta$  parameter of the considered distribution

#### Details :

density function :

$$f(x) = \frac{\exp\left(\frac{x-\alpha}{\beta}\right)}{\beta \left[1 + \exp\left(\frac{x-\alpha}{\beta}\right)\right]^2} \mathbf{1}_{[\alpha, +\infty[}(x)$$

relation between parameters set :

$$\begin{aligned} \mu &= \alpha \\ \sigma &= \sqrt{\frac{1}{3}\pi^2\beta^2} \end{aligned}$$

where

$$\mu = \mathbb{E}[X]$$

$$\sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.



### 2.2.8 LogNormal

This class inherits from the Distribution class.

#### Usage :

Main parameters set :  $LogNormal(\mu_\ell, \sigma_\ell, \gamma)$

Second parameters set :  $LogNormal(\mu, \sigma, \gamma, LogNormal.MUSIGMA)$

Third parameters set :  $LogNormal(\mu, \sigma/\mu, \gamma, LogNormal.MUSIGMAOVERMU)$

Default construction :  $LogNormal()$

#### Arguments :

$\mu_\ell$  : a real value, mean value of  $\log(X)$ ,

$\sigma_\ell$  : a real value, standard deviation of  $\log(X)$ , constraint :  $\sigma_\ell > 0$

$\gamma$  : a real value

$\mu$  : a real value, mean value, constraint :  $\mu > \gamma$

$\sigma$  : a real value, standard deviation, constraint :  $\sigma > 0$

**Value :** a LogNormal . In the default construction, we use the  $LogNormal(mu_\ell, sigma_\ell, gamma) = LogNormal(0.0, 1.0, definition)$ .

#### Some methods :

*getGamma*

**Usage :** *getGamma()*

**Arguments :** none

**Value :** a real value, the  $\gamma$  parameter of the LogNormal distribution

*getMu*

**Usage :** *getMu()*

**Arguments :** none

**Value :** a real value, the  $\mu$  parameter of the LogNormal distribution

*getMuLog*

**Usage :** *getMuLog()*

**Arguments :** none

**Value :** a real value, the  $\mu_\ell$  parameter of the LogNormal distribution

*getSigma*

**Usage :** *getSigma()*

**Arguments :** none

**Value :** a real value, the  $\sigma$  parameter of the LogNormal distribution

*getSigmaLog*

**Usage :** *getSigmaLog()*

**Arguments :** none

**Value :** a real value, the  $\sigma_\ell$  parameter of the LogNormal distribution

*getSigmaOverMu*

**Usage :** *getSigmaOverMu()*

**Arguments :** none

**Value :** a real value, the  $\sigma/\mu$  parameter of the considered distribution

#### Details :

density probability function :

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma_\ell(x - \gamma)} e^{-\frac{1}{2}\left(\frac{\log(x-\gamma)-\mu_\ell}{\sigma_\ell}\right)^2} \mathbf{1}_{[\gamma, +\infty[}(x)$$

relation between parameters set :

$$\begin{aligned} \mu &= e^{\mu_\ell + \sigma_\ell^2/2} + \gamma \\ \sigma &= e^{\mu_\ell + \sigma_\ell^2/2} \sqrt{e^{\sigma_\ell^2} - 1} \end{aligned}$$

where

$$\mu = \mathbb{E}[X]$$

$$\sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.9 MultiNomial

This class inherits from the Distribution class.

**Usage :** Main parameters set : *MultiNomial*( $p, N$ )

**Arguments :**

$p$  : NumericalPoint of dimension  $n$ , constraint :  $0 \leq p_i \leq 1, q = \sum_{i=1}^n p_i \leq 1$

$N$  : an integer,

**Value :** a Multinomial

**Some methods :**

*getN*

**Usage :** *getN*()

**Arguments :** none

**Value :** a integer, the  $N$  parameter of the considered distribution

*getP*

**Usage :** *getP*()

**Arguments :** none

**Value :** a NumericalPoint, the  $p$  parameter of the considered distribution

**Details :**

probability function :

$$P(\underline{X} = \underline{x}) = \frac{N!}{x_1! \dots x_n! (N-s)!} p_1^{x_1} \dots p_n^{x_n} (1-q)^{N-s}$$

with  $0 \leq p_i \leq 1, x_i \in \mathbb{N}, q = \sum_{i=1}^n p_i \leq 1, s = \sum_{i=1}^n x_i \leq N$

relation between parameters set :

$$\begin{aligned} \mu_i &= n p_i \\ \sigma_i &= \sqrt{n p_i (1 - p_i)} \\ \sigma_{i,j} &= -n p_i p_j \end{aligned}$$

$$\text{where} \quad \mu_i = \text{E}[X]_i \quad \sigma_i = \sqrt{\text{Var}[X]_i} \quad \sigma_{i,j} = \text{Cov}[X_i, X_j]$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.10 NonCentralStudent

This class inherits from the Distribution class.

**Usage :**

Main parameters set : *NonCentralStudent*( $\nu, \delta, \gamma$ )

Default construction : *NonCentralStudent*()

**Arguments :**

$\nu$  : a real positive value, generalised number degree of freedom, constraint :  $\nu > 0$

$\delta$  : a real value

$\gamma$  : a real value

**Value :** Student. In the default construction, we use the *NonCentralStudent*(*nu*, *delta*, *gamma*) = *NonCentralStudent* definition.

**Some methods :**

*getMu*

**Usage :** *getMu*()

**Arguments :** none

**Value :** a real value, the  $\mu$  parameter of the NonCentralStudent distribution

*getDelta*

**Usage :** *getDelta*()

**Arguments :** none

**Value :** a real value, the  $\delta$  parameter of the NonCentralStudent distribution

*getGamma*

**Usage :** *getGamma*()

**Arguments :** none

**Value :** a real value, the  $\gamma$  parameter of the NonCentralStudent distribution

**Details :**

density function :

$$f(x) = \frac{1}{\sqrt{\nu} B(\frac{1}{2}, \frac{\nu}{2})} \left( 1 + \frac{(x - \mu)^2}{\nu} \right)^{-\frac{1}{2}(\nu+1)}$$

where  $B$  is the  $\beta$ -function

relation between parameters set :

$$\sigma = \sqrt{\frac{\nu}{\nu - 2}}$$

where

$$\mu = E[X]$$

$$\sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.11 Normal

**Usage :**

*Normal(mean, standardDeviation)*

*Normal(dim)*

*Normal( $\mu$ ,  $\sigma$ ,  $R$ )*

*Normal( $\mu$ ,  $\Sigma$ )*

Default construction : *Normal()*

**Arguments :**

*mean* : a scalar, the mean value of the 1D normal distribution

*standardDeviation* : a scalar, the standard deviation value of the 1D normal distribution

*dim*, an integer : the dimension of the Normal distribution

$\mu$  : a NumericalPoint, the mean of the Distribution

$\sigma$  : a NumericalPoint, the standard deviation of each component, constraint :  $\sigma[i] > 0, \forall i$

$R$  : a CorrelationMatrix, the linear correlation matrix of the Normal distribution

$\Sigma$  : a CovarianceMatrix, the covariance matrix of the Normal distribution

**value :**

while using the first usage, a 1D normal distribution with *mean* as mean value, *standardDeviation* as standard deviation

while using the second usage, a normal distribution of dimension *dim*, with 0 mean value vector, 1-standard deviation vector and identity correlation matrix

while using the third usage, a nD normal distribution with  $\mu$  as mean vector,  $\sigma$  as standard deviation vector and  $R$  as linear correlation matrix

while using the fourth usage, a nD normal distribution with  $\mu$  as mean vector,  $\Sigma$  as covariance matrix

while using the default usage, a 1D normal distribution with 0 mean and unit variance.

**Some methods :**

*getMu*

**Usage :** *getMu()*

**Arguments :** none

**Value :** a NumericalPoint, the  $\mu$  parameter of the distribution

*getSigma*

**Usage :** *getSigma()*

**Arguments :** none

**Value :** a NumericalPoint, the  $\sigma$  parameter of the distribution

*getCorrelationMatrix*

**Usage :** *getCorrelationMatrix()*

**Arguments :** none

**Value :** a CorrelationMatrix, the  $R$  parameter of the distribution

**Details :** probability density function :

$$\frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} e^{-\frac{1}{2}(x-\mu)^t \Sigma^{-1} (x-\mu)}$$

with  $\Sigma = \Lambda(\sigma) R \Lambda(\sigma)$ ,  $\Lambda(\sigma) = \text{diag}(\sigma)$ ,  $R$  symmetric, definite and positive,  $\sigma_i > 0$ .

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.12 Poisson

This class inherits from the Distribution class.

**Usage :** Main parameters set : *Poisson*( $\lambda$ )

**Arguments :**  $\lambda$  : real value, mean and variance value, constraint :  $\lambda > 0$

**Value :** Poisson

**Some methods :**

*getLambda*

**Usage :** *getLambda*()

**Arguments :** none

**Value :** a real positive value, the  $\lambda$  parameter of the considered distribution

**Details :**

probability function :

$$\mathbb{P}(k) = \frac{\lambda^k}{k!} e^{-\lambda}, k \in \mathbb{N}$$

relation between parameters set :

$$\begin{aligned}\mu &= \lambda \\ \sigma &= \sqrt{\lambda}\end{aligned}$$

where

$$\mu = \mathbb{E}[X]$$

$$\sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.13 Student

This class inherits from the Distribution class.

#### Usage :

Main parameters set : *Student*( $\nu, \mu$ )

Default construction : *Student*()

#### Arguments :

$\nu$  : a real positive value, generalised number degree of freedom, constraint :  $\nu > 2$

$\mu$  : a real value, the mean value

**Value :** Student . In the default construction, we use the *Student*( $\nu, \mu$ ) = *Student*(3.0, 0.0) definition.

#### Some methods :

*getMu*

**Usage :** *getMu*()

**Arguments :** none

**Value :** a real value, the  $\mu$  parameter of the Beta distribution

*getNu*

**Usage :** *N*()

**Arguments :** none

**Value :** a real value, the  $\nu$  parameter of the Beta distribution

#### Details :

density function :

$$f(x) = \frac{1}{\sqrt{\nu} B(\frac{1}{2}, \frac{\nu}{2})} \left( 1 + \frac{(x - \mu)^2}{\nu} \right)^{-\frac{1}{2}(\nu+1)}$$

where  $B$  is the  $\beta$ -function and  $\nu \geq 2$

relation between parameters set :

$$\sigma = \sqrt{\frac{\nu}{\nu - 2}}$$

where

$$\mu = E[X]$$

$$\sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.



### 2.2.14 Triangular

This class inherits from the Distribution class.

#### Usage :

Main parameters set :  $\text{Triangular}(a, m, b)$

Default construction :  $\text{Triangular}()$

#### Arguments :

$a$  : a real value, the lower bound

$b$  : a real value, the upper bound, constraint :  $b \geq a$

$m$  : a real value, the mode, constant,  $b \geq m \geq a$

**Value :** Triangular. In the default construction, we use the  $\text{Triangular}(a, m, b) = \text{Triangular}(-1.0, 0.0, 1.0)$  definition.

#### Some methods :

*getA*

**Usage :** *getA()*

**Arguments :** none

**Value :** a real value, the  $a$  parameter of the Triangular distribution

*getB*

**Usage :** *getB()*

**Arguments :** none

**Value :** a real value, the  $b$  parameter of the Triangular distribution

*getM*

**Usage :** *getM()*

**Arguments :** none

**Value :** a real value, the  $m$  parameter of the Triangular distribution

#### Details :

density function :

$$f(x) = \begin{cases} \frac{2(x-a)}{(m-a)(b-a)} & a \leq x \leq m \\ \frac{2(b-x)}{(b-m)(b-a)} & m \leq x \leq b \\ 0 & \text{elsewhere} \end{cases}$$

relation between parameters set :

$$\begin{aligned}\mu &= \frac{1}{3}(a + m + b) \\ \sigma &= \sqrt{\frac{1}{18}(a^2 + b^2 + m^2 - ab - am - bm)}\end{aligned}$$

$$\text{where} \qquad \mu = \text{E}[X] \qquad \sigma = \sqrt{\text{Var}[X]}$$

**Links :** see `docref_B121_ChoixLoi`

Each *getMethod* is associated to a *setMethod*.

### 2.2.15 TruncatedNormal

This class inherits from the Distribution class.

#### Usage :

Main parameters set : *TruncatedNormal*( $\mu_n, \sigma_n, a, b$ )

Default construction : *TruncatedNormal*()

#### Arguments :

$\mu_n$  : a real value which corresponds to the mean of the associated non truncated normal

$\sigma_n$  : a real value which corresponds to the standard deviation of the associated non truncated normal

$a$  : a real value, the lower bound

$b$  : a real value, the upper bound, constraint :  $b \geq a$

**Value :** *TruncatedNormal* . In the default construction, we use the *TruncatedNormal*( $\mu_n, \sigma_n, a, b$ ) = *TruncatedNormal*(0.0, 1.0, -1.0, 1.0) definition.

#### Some methods :

*getA*

**Usage :** *getA*()

**Arguments :** none

**Value :** a real value, the  $a$  parameter of the *TruncatedNormal* distribution

*getB*

**Usage :** *getB*()

**Arguments :** none

**Value :** a real value, the  $b$  parameter of the *TruncatedNormal* distribution

*getMu*

**Usage :** *getMu*()

**Arguments :** none

**Value :** a real value, the  $\mu_n$  parameter of the *TruncatedNormal* distribution

*getSigma*

**Usage :** *getSigma*()

**Arguments :** none

**Value :** a real value, the  $\sigma_n$  parameter of the *TruncatedNormal* distribution

#### Details :

probability density function :

$$f(x) = \frac{\frac{1}{\sigma_n} \phi\left(\frac{x-\mu_n}{\sigma_n}\right)}{\Phi\left(\frac{b-\mu_n}{\sigma_n}\right) - \Phi\left(\frac{a-\mu_n}{\sigma_n}\right)} \mathbf{1}_{[a,b]}(x)$$

(where  $\phi$  and  $\Phi$  are, respectively, the probability density distribution function and the cumulative distribution function of a standard normal distribution)

relation between parameters set :

$$\begin{aligned}\mu &= \mu_n - \sigma_n \frac{\phi(b_{red}) - \phi(a_{red})}{\Phi(b_{red}) - \Phi(a_{red})} \\ \sigma &= \sigma_n \left\{ 1 - \frac{b_{red} \phi(b_{red}) - a_{red} \phi(a_{red})}{\Phi(b_{red}) - \Phi(a_{red})} - \left[ \frac{\phi(b_{red}) - \phi(a_{red})}{\Phi(b_{red}) - \Phi(a_{red})} \right]^2 \right\}^{1/2}\end{aligned}$$

where

$$a_{red} = \frac{a - \mu_n}{\sigma_n} \quad b_{red} = \frac{b - \mu_n}{\sigma_n}$$

and

$$\mu = \mathbb{E}[X] \quad \sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.16 Uniform

This class inherits from the Distribution class.

#### Usage :

Main parameters set :  $Uniform(a, b)$

Default construction :  $Uniform()$

#### Arguments :

$a$  : a real value, the lower bound

$b$  : a real value, the upper bound, constraint :  $b \geq a$

**Value :** Uniform. In the default construction, we use the  $Uniform(a, b) = Uniform(-1.0, 1.0)$  definition.

#### Some methods :

*getA*

**Usage :** *getA()*

**Arguments :** none

**Value :** a real value, the  $a$  parameter of the Uniform distribution

*getB*

**Usage :** *getB()*

**Arguments :** none

**Value :** a real value, the  $b$  parameter of the Uniform distribution

#### Details :

density function :

$$f(x) = \begin{cases} \frac{1}{(b-a)} & a \leq x \leq b \\ 0 & \text{elsewhere} \end{cases}$$

relation between parameters set :

$$\begin{aligned} \mu &= \frac{a+b}{2} \\ \sigma &= \frac{b-a}{2\sqrt{3}} \end{aligned}$$

where

$$\mu = E[X]$$

$$\sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.17 Uniform

This class inherits from the Distribution class.

**Usage :** Main parameters set :  $Uniform(a, b)$

**Arguments :**

$a$  : a real value, the lower bound

$b$  : a real value, the upper bound, constraint :  $b \geq a$

**Value :** TruncatedNormal

**Some methods :**

*getA*

**Usage :** *getA()*

**Arguments :** none

**Value :** a real value, the lower bound

*getB*

**Usage :** *getB()*

**Arguments :** none

**Value :** a real value, the upper bound

**Details :**

probability density function :

$$f(x) = \frac{1}{b-a} \mathbf{1}_{[a,b]}(x)$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.2.18 UserDefined

This class inherits from the Distribution class.

**Usage :** *UserDefined(Coll)*

**Arguments :** *Coll* : a UserDefinedPairCollection, *constraint* : the collection of UserDefinedPair of the UserDefinedPairsCollection must be such that  $\sum_1^n p_i = 1.0$

**Value :** a UserDefined

**Some methods :**

*getPairCollection*

**Usage :** *getPairCollection()*

**Arguments :** none

**Value :** a UserDefinedPairCollection, the *Coll* parameter of the considered distribution

**Details :**

probability function :

$$\mathbb{P}(x_i) = p_i, \quad i = 1, \dots, n$$

where

$(x_i, p_i)$  and  $i = 1, \dots, n$  are respectively a NumericalPoint and its associated probability  
 $n$  is the size of the UserDefinedPairCollection

One must have

$$\sum_{i=1}^n p_i = 1$$

Each *getMethod* is associated to a *setMethod*.

### 2.2.19 Weibull

This class inherits from the Distribution class.

#### Usage :

Main parameters set :  $Weibull(\alpha, \beta, \gamma)$

Second parameter set :  $Weibull(\mu, \sigma, \gamma, 1)$

Default construction :  $Weibull()$

#### Arguments :

$\alpha$  : a real value, the shape parameter, constraint :  $\alpha > 0$

$\beta$  : a real value, the scale parameter, constraint :  $\beta > 0$

$\gamma$  : a real value, the location parameter

$\mu$  : a real value, the mean value,

$\sigma$  : a real value, the standard deviation value, constraint :  $\sigma > 0$

**Value :** a Weibull. In the default construction, we use the  $Weibull(\alpha, \beta, \gamma) = Weibull(1.0, 1.0, 0.0)$  definition.

#### Some methods :

##### *getAlpha*

**Usage :** *getAlpha()*

**Arguments :** none

**Value :** a real value, the  $\alpha$  of the considered distribution

##### *getBeta*

**Usage :** *getBeta()*

**Arguments :** none

**Value :** a real value, the  $\beta$  of the considered distribution

##### *getGamma*

**Usage :** *getGamma()*

**Arguments :** none

**Value :** a real value, the  $\gamma$  parameter of the considered distribution

##### *getMu*

**Usage :** *getMu()*

**Arguments :** none

**Value :** a real value, the  $\mu$  parameter of the considered distribution

##### *getSigma*

**Usage :** *getSigma()*



**Arguments :** none

**Value :** a real value, the  $\sigma$  parameter of the considered distribution

**Details :**

density function :

$$f(x) = \frac{\beta}{\alpha} \left( \frac{x - \gamma}{\alpha} \right)^{\beta-1} \exp \left[ - \left( \frac{x - \gamma}{\alpha} \right)^{\beta} \right] \mathbf{1}_{[\gamma, +\infty[}(x)$$

relation between parameters set :

$$\begin{aligned} \mu &= \alpha \Gamma \left( 1 + \frac{1}{\beta} \right) + \gamma \\ \sigma &= \alpha \sqrt{\Gamma \left( 1 + \frac{2}{\beta} \right) - \Gamma^2 \left( 1 + \frac{1}{\beta} \right)} \end{aligned}$$

where  $\Gamma$  is the  $\Gamma$ -function and

$$\mu = \mathbb{E}[X] \qquad \sigma = \sqrt{\text{Var}[X]}$$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

## 2.3 Truncated distribution

### 2.3.1 TruncatedDistribution

This class enables to truncate any distribution within a specified range which one of the bounds may be infinite. It offers the methods of the Distribution class.

**Usage :**

*TruncatedDistribution(distribution, lowerBound, upperBound)*

*TruncatedDistribution(distribution, bound, TruncatedDistribution.UPPER)*

*TruncatedDistribution(distribution, bound, TruncatedDistribution.LOWER)*

**Arguments :**

*distribution* : a Distribution

*lowerBound* : a real, the new lower bound of the distribution : the distribution range is  $[lowerBound, \infty[$  or  $[lowerBound, max[$  if the distribution is already bounded by *max*

*upperBound* : a real, the new upper bound of the distribution : the distribution range is  $[-\infty, upperBound[$  or  $[min, upperBound[$  if the distribution is already bounded by *min*

**Value :** a Distribution

**Links :** see `docref_B122_Copules_en`

Each *getMethod* is associated to a *setMethod*.

## 2.4 Composed distribution

### 2.4.1 Copula

This class inherits from the `Distribution` class.

**Usage :** *Copula(distributionImplementation)*

**Arguments :** *distributionImplementation* : a `DistributionImplementation`, which must verify the properties of a copula. This distribution can be a `IndependentCopula`, `NormalCopula`, `ClaytonCopula`, `GumbelCopula` or `FrankCopula`.

**Value :** a `Copula`

**Links :** see `docref_B122_Copules_en`

Each *getMethod* is associated to a *setMethod*.

### 2.4.2 ClaytonCopula

This class inherits from the Distribution class.

**Usage :**

*ClaytonCopula()*

*ClaytonCopula(theta)*

**Arguments :** *theta* : a real, the only parameter of the Clayton copula, which PDF is :  $\left(u_1^{-\theta} + u_2^{-\theta} - 1\right)^{-1/\theta}$ ,  
for  $u_i \in [0, 1]$

**Value :**

In the first usage, a ClaytonCopula of dimension 2 with  $\theta = 2.0$ ,

In the second usage, a ClaytonCopula of dimension 2 with the  $\theta$  specified.

**Links :** see docref\_B122\_Copules\_en

### 2.4.3 FrankCopula

This class inherits from the Distribution class.

**Usage :**

*FrankCopula()*

*FrankCopula(theta)*

**Arguments :** *theta* : a real, the only parameter of the Gumbel copula, which PDF is :

$$-\frac{1}{\theta} \log \left( 1 + \frac{(e^{-\theta u_1} - 1)(e^{-\theta u_2} - 1)}{e^{-\theta} - 1} \right)$$

, for  $u_i \in [0, 1]$

**Value :**

In the first usage, a FrankCopula of dimension 2 with  $\theta = 2.0$ ,

In the second usage, a FrankCopula of dimension 2 with the  $\theta$  specified.

**Links :** see docref\_B122\_Copules\_en

### 2.4.4 GumbelCopula

This class inherits from the Distribution class.

**Usage :**

*GumbelCopula()*

*GumbelCopula(theta)*

**Arguments :** *theta* : a real, the only parameter of the Gumbel copula, which PDF is :

$$\exp \left( - \left( (-\log(u_1))^\theta + (-\log(u_2))^\theta \right)^{1/\theta} \right)$$

, for  $u_i \in [0, 1]$

**Value :**

In the first usage, a GumbelCopula of dimension 2 with  $\theta = 2.0$ ,

In the second usage, a GumbelCopula of dimension 2 with the  $\theta$  specified.

**Links :** see docref\_B122\_Copules\_en

### 2.4.5 IndependentCopula

This class inherits from the Distribution class.

**Usage :**

*IndependentCopula()*

*IndependentCopula(dim)*

**Arguments :** *dim* : an integer, the dimension of the copula

**Value :**

In the first usage, a IndependentCopula of dimension 1,

In the second usage, a IndependentCopula of the dimension *dim* specified.

**Links :** see docref\_B122\_Copules\_en

### 2.4.6 NormalCopula

This class inherits from the Distribution class.

**Usage :**

*NormalCopula()*

*NormalCopula(R)*

**Arguments :** *R* : a CorrelationMatrix which is not the Kendall nor the Spearman rank correlation matrix of the distribution. The *R* matrix can be evaluated from the Spearman or Kendall correlation matrix

**Value :**

In the first usage, a NormalCopula of dimension 1

In the second usage, a NormalCopula with the correlation matrix *R* specified.

**Some methods :**

*getNormalCorrelationFromKendallCorrelation*

**Usage :** *NormalCopula.getNormalCorrelationFromKendallCorrelation(K)*

**Arguments :** *K* : a CorrelationMatrix, it must be the Kendall correlation matrix of the considered random vector

**Value :** a CorrelationMatrix, the correlation matrix of the normal copula evaluated from the Kendall correlation matrix *K*

*getNormalCorrelationFromSpearmanCorrelation*

**Usage :** *NormalCopula.getNormalCorrelationFromSpearmanCorrelation(S)*

**Arguments :** *S* : a CorrelationMatrix, it must be the Spearman correlation matrix of the considered random vector

**Value :** a CorrelationMatrix, the correlation matrix of the normal copula evaluated from the Spearman correlation matrix *S*

**Links :** see [docref\\_B122\\_Copules\\_en](#)



### 2.4.7 ComposedCopula

This class inherits from the Distribution class.

**Usage :** *ComposedCopula(copulaCollection)*

**Arguments :** *copulaCollection* : a CopulaCollection

**Value :** a ComposedCopula, defined as the product of the initial copulas. For example, if  $C_1$  and  $C_2$  are two copulas respectively of  $\mathcal{R}^{n_1}$  and  $\mathcal{R}^{n_2}$ , we can create the copula of a random vector of  $\mathcal{R}^{n_1+n_2}$ , noted  $C$  as follows :

$$C(u_1, \dots, u_n) = C_1(u_1, \dots, u_{n_1})C_2(u_{n_1+1}, \dots, u_{n_1+n_2})$$

It means that both subvectors  $(u_1, \dots, u_{n_1})$  and  $(u_{n_1+1}, \dots, u_{n_1+n_2})$  of  $\mathcal{R}^{n_1}$  and  $\mathcal{R}^{n_2}$  are independent.

**Some methods :**

*getCopulaCollection*

**Usage :** *getCopulaCollection()*

**Arguments :** none

**Value :** a CopulaCollection, the collection of copulas from which the ComposedCopula is built

**Links :** see docref\_B\_JoinedCDF\_en

### 2.4.8 ComposedDistribution

This class inherits from the `Distribution` class.

**Usage :** *ComposedDistribution(distributionCollection, copula)*

**Arguments :**

*distributionCollection* : a `DistributionCollection`

*copula* : a `Copula`

**Value :** a `ComposedDistribution`

**Some methods :**

*getDistributionCollection*

**Usage :** *getDistributionCollection()*

**Arguments :** none

**Value :** a `DistributionCollection`, the collection of distributions from which the `ComposedDistribution` is built

**Links :** see `docref_B_JoinedCDF_en`

Each *getMethod* is associated to a *setMethod*.

## 2.5 Mixture

### 2.5.1 Mixture

A Mixture is a distribution such that its probability density function is a linear combination of probability density functions, with the linear combination coefficients greater or equal to zero such that their sum is equal to 1.

It is important to note that the linear combination coefficients are given through the *weight* attribute of each component of the *DistributionCollection*, thanks to the command *DistributionCollection[i].setWeight(coefficient)*.

**Usage :** *Mixture(distributionCollection)*

**Arguments :** *distributionCollection* : a *DistributionCollection*, the collection of the distributions which composes the linear combination

**Value :** a Mixture

**Some methods :**

*getDistributionCollection*

**Usage :** *getDistributionCollection()*

**Arguments :** none

**Value :** a *DistributionCollection* the collection of distribution from which the Mixture is built

**Details :**

probability density function :

$$f(x) = \sum \alpha_i p_i(x)$$

with  $\sum \alpha_i = 1$

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

## 2.6 KernelMixture

### 2.6.1 KernelMixture

A KernelMixture is a distribution built from a NumericalSample, such that its probability density function is a linear combination of the kernel specified by the User, centered on each point of the NumericalSample, which standard deviation is the bandwidth specified by the User. It is important to note that the linear combination coefficients are all equal.

**Usage :** *KernelMixture(kernel, bandwidth, sample)*

**Arguments :** *distributionCollection* : a DistributionCollection, the collection of the distributions which composes the linear combination

**Value :** a KernelMixture

**Some methods :**

*getBandwidth*

**Usage :** *getBandwidth()*

**Arguments :** none

**Value :** a NumericalPoint, the bandwidth of the kernel mixture, (see equation below for dimension 1). The bandwidth is the same at each point of the NumericalSample

*getKernel*

**Usage :** *getKernel()*

**Arguments :** none

**Value :** a Distribution, the kernel  $K$  of the mixture, (see equation below for dimension 1)

*getSample*

**Usage :** *getSample()*

**Arguments :** none

**Value :** a NumericalSample, the NumericalSample of the mixture, (see equation below for dimension 1)

**Details :**

Probability density function in dimension 1 :

$$f(x) = \sum \frac{1}{nh} K\left(\frac{X_i - x}{h}\right), x \in \mathbb{R}$$

where  $(X_1, \dots, X_n)$  is a NumericalSample

**Links :** see docref\_B121\_ChoixLoi

Each *getMethod* is associated to a *setMethod*.

### 2.6.2 KernelSmoothing

The class KernelSmoothing enables to build some kernels used to fit a distribution to a numerical sample.

**Usage :**

```
KernelSmoothing()
KernelSmoothing(Distribution(myDistribution))
KernelSmoothing(DistributionImplementation())
```

**Arguments :**

*myDistribution* : a 1D Distribution of any kind  
*DistributionImplementation()* : default constructor of the 1D UsualDistribution. For example, *Uniform()*, *Triangular()*, ...

**Value :** a Distribution

In the first usage, the kernel is the kernel product of 1D Normal(1.0, 0.0). The dimension of the product is detected from the numerical sample.

In the second usage, the kernel is the kernel product of 1D distributions specified by *myDistribution*. Care : the kernel smoothing method is all the more efficient than the kernel is symmetric with respect to 0.0. The dimension of the product is detected from the numerical sample.

In the third usage, the kernel is the kernel product of the default constructions of the 1D UsualDistributions. Note that the default constructor of a UsualDistribution builds a distribution which is symmetric with respect to 0.0 when it is possible. The dimension of the product is detected from the numerical sample.

**Some methods :**

*buildImplementation*

**Usage :**

```
buildImplementation(sample)
buildImplementation(sample, boundaryCorrection)
buildImplementation(sample, bandwidth)
buildImplementation(sample, bandwidth, boundaryCorrection)
```

**Arguments :**

*sample* : a NumericalSample, the numerical sample from which the kernel mixture is built  
*boundaryCorrection* : a Bool which indicates if it is necessary to make a boundary treatment (according to the mirroring technique)  
*bandwidth* : a NumericalPoint, the bandwidth of the kernel product. The dimension is detected from the numerical sample *sample* and evaluated according to the Scott rule.

**Value :** a Distribution

*computeSilvermanBandwidth*

**Usage :** *computeSilvermanBandwidth(sample)*

**Arguments :** *sample* : a NumericalSample, the numerical sample from which the kernel mixture is built

**Value :** a NumericalPoint, the bandwidth automatically evaluated by Open TURNS from the numerical sample according to the Silverman rule

*getBandwidth*

**Usage :** *getBandwidth()*

**Arguments :** none

**Value :** a NumericalPoint, the bandwidth of the kernel mixture, (see equation below for dimension 1). The bandwidth is the same at each point of the NumericalSample

*getKernel*

**Usage :** *getKernel()*

**Arguments :** none

**Value :** a Distribution, the kernel adopted for the kernel smoothing

#### Details :

Probability density function in dimension 1 :

$$p_n(x) = \sum \frac{1}{nh} K\left(\frac{x - X^i}{h}\right), x \in \mathbb{R}$$

where  $(X^1, \dots, X^n)$  is a NumericalSample and  $K$  the kernel PDF,  
Probability density function in dimension  $N$  :

$$p_n(\vec{x}) = \frac{1}{n} \sum_{i=1}^{i=N} \prod_{j=1}^{j=N} \frac{1}{h_j} K\left(\frac{x^j - X_i^j}{h_j}\right)$$

where  $\prod_{j=1}^{j=N} K(x^j)$  is the kernel product and  $\underline{h} = (h^1, \dots, h^N)$  the vector of bandwidth.

**Links :** see `docref_B121_ChoixLoi`

## 2.7 Random vector

### 2.7.1 RandomVector

Usage :

*RandomVector(distribution)*

*RandomVector(function, distribution)*

Arguments :

*distribution* : a Distribution

*function* : a NumericalMathFunction

Value : a RandomVector, which is of type :

*Usual* : if created thanks to the first usage. In that case, the RandomVector has for distribution the one specified through *distribution*

*Composite* : if created thanks to the second usage. In that case, the RandomVector is defined as the function of a RandomVector  $X$  which distribution is *distribution* :  $Y = function(X)$

Some methods :

*getAntecedent*

Usage : *getAntecedent()*

Arguments : no argument

Value : a RandomVector, only in the case of Composite RandomVector : the RandomVector  $X$  such that  $Y = function(X)$ .

*getCovariance*

Usage : *getCovariance()*

Arguments : no argument

Value : a CovarianceMatrix, only in the case of Usual RandomVector : the covariance of the considered RandomVector

*getDistribution*

Usage : *getDistribution()*

Arguments : no argument

Value : a Distribution, only in the case of Usual RandomVector : the distribution of the RandomVector

*getDescription*

Usage : *getDescription()*

Arguments : no argument

Value : a Description, the description of the Randomvector

*getDimension*

**Usage :** *getDimension()*

**Arguments :** no argument

**Value :** an integer, the dimension of the RandomVector

*getMarginal*

**Usage :**

*getMarginal(i)*

*getMarginal(indices)*

**Arguments :**

*i* : an integer which indicates the component concerned

*indices* : a Indices which regroups all the components concerned

**Value :** a RandomVector restricted to the concerned components.

**Details :** Let's note  $\underline{Y} = (Y_1, \dots, Y_n)^t$  a random vector and  $I \in [1, n]$  a set of indices. If  $\underline{Y}$  is a UsualRandomvector, the subvector is defined by  $\tilde{\underline{Y}} = (Y_i)_{i \in I}^t$ . If  $\underline{Y}$  is a CompositeRandomVector, defined by  $\underline{Y} = f(\underline{X})$  with  $f = (f_1, \dots, f_n)$ ,  $f_i$  some scalar functions, the sub vector is  $\tilde{\underline{Y}} = (f_i(\underline{X}))_{i \in I}$ .

*getMean*

**Usage :** ()

**Arguments :** no argument

**Value :** a NumericalPoint, only in the case of Usual RandomVector : the mean vector of the associated distribution

*getName*

**Usage :** *getName()*

**Arguments :** no argument

**Value :** a string, the name of the RandomVector

*getNumericalSample*

**Usage :** *getNumericalSample()*

**Arguments :** no argument

**Value :** a NumericalSample a sample of the random vector

*isComposite*

**Usage :** *isComposite()*

**Arguments :** no argument

**Value :** a boolean which indicates if the RandomVector is of type Composite or Usual.

Each *getMethod* is associated to a *setMethod*.



## 3 Experiment planes

### 3.1 Experiment

**Usage :** *Experiment*(*expPlaneImplentation*)

**Arguments :** *expPlaneImplentation* : an ExperimentImplementation, which is Axial, Factorial, Composite or Box

**Value :** an Experiment

**Some methods :**

*generate*

**Usage :** *generate*()

**Arguments :** none

**Value :** a NumericalSample, the points which constitute the experiment plane

*getLevels*

**Usage :** *getLevels*()

**Arguments :** none

**Value :** a NumericalPoint, the levels of the experiment of the plane

*getCenter*

**Usage :** *getCenter*()

**Arguments :** none

**Value :** a NumericalPoint, the center of the experiment plane

*.str()*

**Usage :** *str*()

**Arguments :** no argument

**Value :** a string which describes the Experiment

**Links** see `docref_C11_MinMaxPlanExp`

Each *getMethod* is associated to a *setMethod*.

## 3.2 Axial

This class inherits from the Experiment class.

### Usage :

*Axial(center, levels)*

*Axial(dimension, levels)*

### Arguments :

*center* : a NumericalPoint, the center of the experiment plane

*levels* : a NumericalPoint, the levels of the experiment of the plane

*dimension* : an integer, the dimension of the space where the experiment plane is created

### Value : a Axial

if defined with the first usage, the experiment plane is centered on *center*

if defined with the second usage, the experiment plane is centered on the *center* = 0

### Details :

Number of points generated :  $1 + 2 * levels.getDimension() * dimension$

The axial plane generates a NumericalSample where :

the first point is the vector (*center*),

the following points are : each coordinate one at a time is equal to +/- levels[i], for each direction so on, until the last level.

It is possible to use the *scale*, *translate* methods of the NumericalSample in order to scale each direction and translate the grid structure onto the right center

**Links** see docref\_C11\_MinMaxPlanExp

### 3.3 Factorial

This class inherits from the Experiment class.

#### Usage :

*Factorial(center, level)*

*Factorial(dimension, level)*

#### Arguments :

*center* : a NumericalPoint, the center of the experiment plane

*level* : a NumericalPoint, the levels of the experiment of the plane

*dimension* : an integer, the dimension of the space where the experiment plane is created

#### Value : a Factorial

if defined with the first usage, the experiment plane is centered on *center*

if defined with the second usage, the experiment plane is centered on the *center* = 0

#### Details :

Number of points generated :  $1 + levels.getDimension() * 2^{dimension}$

The factorial plane generates a NumericalSample where :

the first point is the vector (*center*),

the following points are : all coordinates are equal to +/- levels[i] for each direction

It is possible to use the *scale*, *translate* methods of the NumericalSample in order to scale each direction and translate the grid structure onto the right center

**Links** see docref\_C11\_MinMaxPlanExp

### 3.4 Composite

This class inherits from the Experiment class.

#### Usage :

*Composite(center, level)*

*Composite(dimension, level)*

#### Arguments :

*center* : a NumericalPoint, the center of the experiment plane

*level* : a NumericalPoint, the levels of the experiment of the plane

*dimension* : an integer, the dimension of the space where the experiment plane is created

#### Value : a Composite

if defined with the first usage, the experiment plane is centered on *center*

if defined with the second usage, the experiment plane is centered on the *center* = 0

#### Details :

A composite plane is the union of an axial and a factorial one

Number of points generated :  $1 + levels.getDimension() * (2 * dimension + 2^{dimension})$

The composite plane generates a NumericalSample where :

the first point is the vector (*center*),

the following points are : one coordinate at a time is equal to +/- levels[i] for each direction

It is possible to use the *scale*, *translate* methods of the NumericalSample in order to scale each direction and translate the grid structure onto the right center

#### Links see docref\_C11\_MinMaxPlanExp

### 3.5 Box

This class inherits from the Experiment class.

#### Usage :

*Box(levelsBox)*

#### Arguments :

*levelsBox* : a NumericalPoint, which specifies the number of intermediate points on each direction (one per direction) which regularly discretize the unit pavement  $[-0.5, 0.5]^n$

**Value :** a Box, which regularly discretizes the unit pavement  $[-0.5, 0.5]^n$  with the specified number of intermediate points for each direction

#### Details :

Number of points generated :  $\prod_{i=1}^n (2 + n_{level}(\text{direction } i))$

The box plane generates a NumericalSample where :

It is possible to use the *scale*, *translate* methods of the NumericalSample in order to scale each direction and translate the grid structure onto the right center

**Links** see docref\_C11\_MinMaxPlanExp

## 4 Statistics on sample

### 4.1 Numerical Sample

#### 4.1.1 NumericalSample

**Usage :**

*NumericalSample(size, dim)*  
*NumericalSample(size, numericalPoint)*

**Arguments :**

*size* : an integer, the size of the NumericalSample  
*dim* : an integer, the dimension each NumericalPoint of the NumericalSample  
*numericalPoint* : a NumericalPoint

**Value :** a NumericalSample, containing *size* NumericalPoint, each one equal to :

$\underline{0} \in \mathbb{R}^{dim}$  in the first usage  
*numericalPoint* in the second usage

**Some methods :**

[]

**Usage :** *NumericalSample[i]*

**Arguments :** *i* : an integer, constraint :  $0 \leq i \leq size - 1$

**Value :** a NumericalPoint, the *i*th NumericalPoint of the NumericalSample

*add*

**Usage :** *add(x)*

**Arguments :** *x* : a NumericalPoint

**Value :** a NumericalSample, of size *size* + 1 where the last NumericalPoint has been added, equal to *x*

*computeCovariance*

**Usage :** *computeCovariance()*

**Arguments :** none

**Value :** a CovarianceMatrix, the covariance matrix of the NumericalSample (a  $dim^2$  matrix)

*computeEmpiricalCDF*

**Usage :** *computeEmpiricalCDF(x)*

**Arguments :** *x* : a NumericalPoint

**Value :** a numerical scalar, the Empirical Cumulative Distribution Function value of the NumericalSample at  $x$

*computeKendallTau*

**Usage :** *computeKendallTau()*

**Arguments :** none

**Value :** a CorrelationMatrix, the Kendall rank correlation matrix of the NumericalSample

*computeKurtosisPerComponent*

**Usage :** *computeKurtosisPerComponent()*

**Arguments :** none

**Value :** a NumericalPoint, the value of the kurtosis of each component of the NumericalSample

*computeMean*

**Usage :** *computeMean()*

**Arguments :** none

**Value :** a NumericalPoint, the mean value vector of each component of the NumericalSample

*computeMedianPerComponent*

**Usage :** *computeMedianPerComponent()*

**Arguments :** none

**Value :** a NumericalPoint, the median value vector of each component of the NumericalSample

*computePearsonCorrelation*

**Usage :** *computePearsonCorrelation()*

**Arguments :** none

**Value :** a CorrelationMatrix, the Pearson correlation matrix of the NumericalSample (a  $dim^2$  matrix)

*computeQuantile*

**Usage :** *computeQuantile(p)*

**Arguments :**  $p$ , a real value, constraint  $0 \leq p \leq 1$ , the value of a probability

**Value :** a NumericalPoint, the empirical quantile value associated to probability  $p$ , determined from the empirical CDF of the NumericalSample

*computeQuantilePerComponent*

**Usage :** *computeQuantilePerComponent(p)*

**Arguments :**  $p$ , a real value, constraint  $0 \leq p \leq 1$ , the value of a probability

**Value :** a NumericalPoint, the empirical quantile value associated to probability  $p$  for each component, determined from the empirical CDF of each component of the NumericalSample

*computeSkewnessPerComponent*

**Usage :** *computeSkewnessPerComponent()*

**Arguments :** none

**Value :** a NumericalPoint, the skewness of each component of the NumericalSample

*computeSpearmanCorrelation*

**Usage :** *computeSpearmanCorrelation()*

**Arguments :** none

**Value :** a CorrelationMatrix, the Spearman correlation matrix of the NumericalSample (a  $dim^2$  matrix)

*computeStandardDeviation*

**Usage :** *computeStandardDeviation()*

**Arguments :** none

**Value :** a SquareMatrix, the Cholesky factor  $\underline{\underline{L}}$  of the covariance matrix  $\underline{\underline{\Lambda}}$  :  $\underline{\underline{L}}\underline{\underline{L}}^t = \underline{\underline{\Lambda}}$ , with  $\underline{\underline{L}}$  triangular inferior

*computeStandardDeviationPerComponent*

**Usage :** *computeStandardDeviationPerComponent()*

**Arguments :** none

**Value :** a NumericalPoint, the standard Deviation value of each component of the NumericalSample

*computeVariancePerComponent*

**Usage :** *computeVariancePerComponent()*

**Arguments :** none

**Value :** a NumericalPoint, the variance value of each component of the NumericalSample

*getDimension*

**Usage :** *getDimension()*

**Arguments :** none

**Value :** an integer, the dimension of each point which constitutes the NumericalSample (it returns *dim*)

*getMin*

**Usage :** *getMin()*

**Arguments :** none

**Value :** a NumericalPoint, each element of the NumericalPoint corresponds to the minimum of each component of the NumericalSample

*getMax*

**Usage :** *getMax()*

**Arguments :** none



**Value :** a NumericalPoint, each element of the NumericalPoint corresponds to the maximum of each component of the NumericalSample

### *getMarginal*

**Usage :**

*getMarginal(i)*

*getMarginal(i)*

**Arguments :**

*i* : a UnsignedLong, (integer)

*indices* : a Indices (table of intergers)

**Value :**

a NumericalSample : the NumericalSample of same size as the initial NumericalSample, of dimension 1, corresponding to the  $i + 1$  coordinate of the NumericalPoints which constitute the initial NumericalSample

a NumericalSample : the NumericalSample of same size as the initial NumericalSample, of dimension *indices.getSize()*, corresponding to the associated coordinates of the NumericalPoints which constitute the initial NumericalSample. Care : indices are initialized to 0.

### *getSize*

**Usage :** *getSize()*

**Arguments :** none

**Value :** an integer, the size of the NumericalSample (which means the number of points which constitute the NumericalSample (it returns *size*)

### *rank*

**Usage :** *rank()*

**Arguments :** none

**Value :** a NumericalSample, where each value has been replaced by the value of its rank (order set component by component)

### *scale*

**Usage :** *scale(factor)*

**Arguments :** *factor* : a NumericalPoint

**Value :** a NumericalSample, where the components [i] of each NumericalPoint have been multiplied by the corresponding value *factor[i]*

### *sort*

**Usage :** *sort(i)*

**Arguments :** *i* : UnsignedLong (an integer)

**Value :** a NumericalSample of same size as the initial NumericalSample, of dimension 1, constituted by the  $i + 1$ th component of the NumericalPoints that constitute the initial NumericalSample, all sorted in ascending order

*sortAccordingAComponent*

**Usage :** *sortAccordingAComponent(i)*

**Arguments :** *i* : UnsignedLong (an integer)

**Value :** a NumericalSample of same size and dimension as the initial NumericalSample, where the NumericalPoints have been reordered such that the  $(i + 1)$  component is sorted in ascending order

*split*

**Usage :** *split(i)*

**Arguments :** *i* : UnsignedLong (an integer)

**Value :** a NumericalSample which contains only the NumericalPoints of the initial NumericalSample corresponding to the indices  $k$ , where  $k \geq i$ . Care : The initial NumericalSample is modified and only contains the first  $i$  NumericalPoints (which means the ones corresponding to the indices  $k$ , where  $k < i$ ).

*str*

**Usage :** *str()*

**Arguments :** none

**Value :** a string giving a brief description of the considered NumericalSample,

*translate*

**Usage :** *translate(translation)*

**Arguments :** *translation* : a NumericalPoint

**Value :** a NumericalSample, where the components  $[i]$  of each NumericalPoint have been added the corresponding value *translation* $[i]$

**Details :**

when two elements of the sample are equal, the rank of the first element appearing in the sample will be considered as the lower one (for computing Spearman correlation matrix)

## 4.2 Distribution factory

### 4.2.1 DistributionImplementationFactory

**Usage :** *DistributionImplementationFactory()*

**Arguments :** none

**Value :** a *DistributionImplementationFactory* is the implementation of a *Factory* and is one of the following classes : *BetaFactory*, *ExponentialFactory*, *GammaFactory*, *GeometricFactory*, *GumbelFactory*, *HistogramFactory*, *LogisticFactory*, *LogNormalFactory*, *MultiNomialFactory*, *NormalFactory*, *PoissonFactory*, *StudentFactory*, *TriangularFactory*, *TruncatedNormalFactory*, *UserDefinedFactory*, *UniformFactory*, *WeibullFactory*.

**Some methods :**

*buildImplementation*

**Usage :** *buildImplementation(sample)*

**Arguments :** *sample* : a *NumericalSample*, of dimension  $n \geq 1$

**Value :** a *DistributionImplementation*, which is a *Beta*, *Exponential*, *Gamma*, *Geometric*, *Gumbel*, *Histogram*, *Logistic*, *LogNormal*, *MultiNomial*, *Normal*, *Poisson*, *Student*, *Triangular*, *TruncatedNormal*, *UserDefined*, *Uniform*, *Weibull*.

### 4.3 Correlation analysis

#### 4.3.1 CorrelationAnalysis

**Usage :** *CorrelationAnalysis()*

**Arguments :** none

**Some methods :**

*PCC*

**Usage :** *PCC(sample1, sample2)*

**Arguments :**

*sample1* : a NumericalSample, of dimension  $n \geq 1$

*sample2* : a NumericalSample, of dimension =1

**Value :** a NumericalPoint, the PCC (partial Pearson Correlation Coefficient) coefficients evaluated between the *sample2* and each coordinate of *sample1*

*PRCC*

**Usage :** *PRCC(sample1, sample2)*

**Arguments :**

*sample1* : a NumericalSample, of dimension  $n \geq 1$

*sample2* : a NumericalSample, of dimension =1

**Value :** a NumericalPoint, the PRCC (Pearson Rank Correlation Coefficient) coefficients evaluated between the *sample2* and each coordinate of *sample1* (based on the rank values)

*PearsonCorrelation*

**Usage :** *PearsonCorrelation(sample1, sample2)*

**Arguments :**

*sample1* : a NumericalSample, of dimension = 1

*sample2* : a NumericalSample, of dimension =1

**Value :** a real value, the Pearson Correlation coefficient evaluated between the *sample2* and *sample1*

*SRC*

**Usage :** *SRC(sample1, sample2)*

**Arguments :**

*sample1* : a NumericalSample, of dimension  $n \geq 1$

*sample2* : a NumericalSample, of dimension =1

**Value :** a NumericalPoint, the SRC (Standard Regression Coefficient) coefficients evaluated between the *sample2* and each coordinate of *sample1*

*SRCC*

**Usage :** *PRCC(sample1, sample2)*

**Arguments :**

*sample1* : a NumericalSample, of dimension  $n \geq 1$

*sample2* : a NumericalSample, of dimension =1

**Value** : a NumericalPoint, the SRCC (Standard Rank Regression Coefficient) coefficients evaluated between the *sample2* and each coordinate of *sample1* (based on the rank values)

### *SpearmanCorrelation*

**Usage** : *SpearmanCorrelation(sample1, sample2)*

**Arguments** :

*sample1* : a NumericalSample, of dimension =1

*sample2* : a NumericalSample, of dimension =1

**Value** : a real value, the Spearman Correlation coefficient evaluated between the *sample2* and *sample1* (based on the rank values)

## 4.4 Fitting test

### 4.4.1 TestResult

**Usage :** a TestResult is the result of a fitting test, of type NormalityTest or HypothesisTest.

**Some methods :**

*getBinaryQualityMeasure*

**Usage :** *getBinaryQualityMeasure()*

**Arguments :** none

**Value :** a boolean value, indicating the succes of the test : 1 for succes and 0 for failure

*getPValue*

**Usage :** *getPValue()*

**Arguments :** none

**Value :** a real positive value,  $<1$ , the p-value of the test

*getTestType*

**Usage :** *getTestType()*

**Arguments :** none

**Value :** a string describing the type of the test

*getThreshold*

**Usage :** *getThreshold()*

**Arguments :** none

**Value :** a real positive value,  $<1$ , the p-value threshold

#### 4.4.2 FittingTest

This class is used through its static methods in order to evaluate some hypothesis on samples : independence or monotonous relation.

##### Some methods :

###### *BestModelBIC*

###### Usage :

*FittingTest().BestModelBIC(sample, factoryCollection)*

*FittingTest().BestModelBIC(sample, distributionCollection)*

###### Arguments :

*sample* : a NumericalSample, the sample which will be tested

*factoryCollection* : a FactoryCollection, the collection of factories which are the structures which build the distribution from a sample

*distributionCollection* : a DistributionCollection, a collection of the distributions which will be tested through the BIC criteria

**Value** : a Distribution, the best one according to the BIC criteria

###### *BestModelChiSquared*

###### Usage :

*FittingTest().BestModelChiSquared(sample, factoryCollection)*

*FittingTest().BestModelChiSquared(sample, distributionCollection)*

###### Arguments :

*sample* : a NumericalSample, the sample which will be tested

*factoryCollection* : a FactoryCollection, a collection of the factories which are the structures which build distributions from a sample

*distributionCollection* : a DistributionCollection, the collection of the distributions which will be tested through the Chi Squared criteria

**Value** : a Distribution, the best one according to the ChiSquared criteria

###### *BestModelKolmogorov*

###### Usage :

*FittingTest().BestModelKolmogorov(sample, factoryCollection)*

*FittingTest().BestModelKolmogorov(sample, distributionCollection)*

###### Arguments :

*sample* : a NumericalSample, the sample which will be tested

*factoryCollection* : a FactoryCollection, a collection of the factories which are the structures which build distributions from a sample

*distributionCollection* : a DistributionCollection, the collection of the distributions which will be tested through the Kolmogorov criteria

**Value** : a Distribution, the best one according to the Kolmogorov criteria

###### *BIC*

**Usage :**

*FittingTest().BIC(sample, factory)*  
*FittingTest().BIC(sample, distribution)*

**Arguments :**

*sample* : a NumericalSample, the sample which will be tested  
*factory* : a Factory, the structure which builds the distribution from the sample which will be tested  
*distribution* : a Distribution, which will be tested through the BIC criteria

**Value :** a real value, the BIC value of the distribution tested evaluated on the sample

*ChiSquared***Usage :**

*FittingTest().ChiSquared(sample, factory, level)*  
*FittingTest().ChiSquared(sample, distribution, level)*

**Arguments :**

*sample* : a NumericalSample, the sample which will be tested  
*factory* : a Factory, the structure which builds the distribution from the sample which will be tested  
*distribution* : a Distribution, which will be tested through the ChiSquared criteria  
*level* : a real value, constraint :  $0 < level < 1$ , such as  $1 - level$  be the first type error of the fitting test (the probability you reject the distribution tested whereas you should not have). If not fulfilled, by default,  $level = 0.95$ .

**Value :** a TestResult, the structure which contains the result of the ChiSquared Test : the first usage tests a type of distribution, the second one tests a particular distribution

*Kolmogorov***Usage :**

*FittingTest().Kolmogorov(sample, factory, level)*  
*FittingTest().Kolmogorov(sample, distribution, level)*

**Arguments :**

*sample* : a NumericalSample, the sample which will be tested  
*factory* : a Factory, the structure which builds the distribution from the sample which will be tested  
*distribution* : a Distribution, which will be tested through the Kolmogorov criteria  
*level* : a real value, constraint :  $0 < level < 1$ , such as  $1 - level$  be the first type error of the fitting test (the probability you reject the distribution tested whereas you should not have). If not fulfilled, by default,  $level = 0.95$ .

**Value :** a TestResult, the structure which contains the result of the Kolmogorov Test : the first usage tests a type of distribution, the second one tests a particular distribution



#### 4.4.3 VisualTest

This class is used through its static methods in order to graphically evaluate some hypothesis on samples : independence or monotonous relation.

##### Some methods :

###### *DrawClouds*

###### Usage :

*VisualTest().DrawClouds(sample1, dist)*

*VisualTest().DrawClouds(sample1, sample2)*

###### Arguments :

*sample1* : a NumericalSample, drawn on the graph

*sample2* : a NumericalSample, drawn on the graph

*dist* : a Distribution, the distribution which pdf is drawn

**Value** : a Graph, the structure wich contains : one curve (pdf) and a cloud for the first usage or two clouds in the second usage

###### *DrawEmpiricalCDF*

**Usage** : *VisualTest().DrawEmpiricalCDF(sample, xMin, xMax)*

###### Arguments :

*sample* : a NumericalSample, drawn on the graph

*xMin* : a real value, the lower boundary of the graph

*xMax* : a real value, the upper boundary of the graph, must be  $> xMin$

**Value** : a Graph, the structure wich contains : one staircase curve which is the empirical cdf of the sample

###### *DrawHenryLine*

**Usage** : *VisualTest().DrawHenryLine(sample)*

**Arguments** : *sample* : a NumericalSample, drawn on the graph

**Value** : a Graph, the structure wich contains one histogram

###### *DrawHistogram*

###### Usage :

*VisualTest().DrawHistogram(sample, barNumber)*

*VisualTest().DrawHistogram(sample)*

###### Arguments :

*sample* : a NumericalSample, which histogram is drawn

*barNumber* : an integer, the number of barplots used to draw the histogram. If not mentioned, the number of barplots is automatically determined by Open TURNS according to the Silverman rule

**Value** : a Graph, the structure wich contains the graph : one curve and a cloud for the first usage or two clouds in the second usage

*DrawLMResidualTest*

**Usage :** *VisualTest().DrawLMResidualTest(sample1, sample2, linearModel)*

**Arguments :**

*sample1* : a NumericalSample,  $X$ , of dimension 1

*sample2* : a NumericalSample,  $Y$ , of dimension 1, which is scalar described by a linear model from  $X : Y = aX + b$ ,  $a$  and  $b$  real values

*linearModel* : a LinearModel, the regression model

**Value :** a Graph, the structure wich contains the cloud of the residual values, couples (residual  $i$ , residual  $i+1$ )

*DrawLMVisualTest*

**Usage :** *VisualTest().DrawLMVisualTest(sample1, sample2, linearModel)*

**Arguments :**

*sample1* : a NumericalSample,  $X$ , of dimension 1

*sample2* : a NumericalSample,  $Y$ , of dimension 1, which is described by a linear model from  $X : Y = aX + b$ ,  $a$  and  $b$  real values

*linearModel* : a LinearModel, the regression model

**Value :** a Graph, the structure wich contains the cloud of points  $(X_i, Y_i)$  and the linear model line  $Y = aX + b$

*DrawQQplot***Usage :**

*VisualTest().DrawQQplot(sample1, sample2)*

*VisualTest().DrawQQplot(sample1, sample2, pointNumber)*

*VisualTest().DrawQQplot(sample1, distribution)*

*VisualTest().DrawQQplot(sample1, distribution, pointNumber)*

**Arguments :**

*sample1* : a NumericalSample, used to build on the graph

*sample2* : a NumericalSample, used to build the graph

*distribution* : a Distribution, the distribution which pdf is drawn

*pointNumber* : an integer, the number of points used to build the graph, equal to 20 by default

**Value :** a Graph, the structure wich contains the corresponding empirical fractiles between the two samples in the two first usages, or between the sample and the distribution in the two last usages

#### 4.4.4 NormalityTest

This class is used through its static methods in order to evaluate whether the sample follows a normal distribution. These two tests give more importance to extreme values.

##### Some methods :

###### *AndersonDarlingNormal*

**Usage :**

*NormalityTest().AndersonDarlingNormal(sample)*

*NormalityTest().AndersonDarlingNormal(sample, level)*

**Arguments :**

*sample* : a NumericalSample, of dimension 1 : the sample tested

*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be < 1, equal to 0.95 by default

**Value :** a TestResult, the structure which contains the result of the test.

**Details :** the AndersonDarlingNormal Test is used to check whether the sample follows a normal distribution. This test gives more importance to extreme values

###### *CramerVonMisesNormal*

**Usage :**

*NormalityTest().CramerVonMisesNormal(sample)*

*NormalityTest().CramerVonMisesNormal(sample, level)*

**Arguments :**

*sample* : a NumericalSample, of dimension 1 : the sample tested

*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be < 1, equal to 0.95 by default

**Value :** a TestResult, the structure which contains the result of the test.

**Details :** the CramerVonMisesNormal Test is used to check whether the sample follows a normal distribution. This test gives more importance to extreme values

#### 4.4.5 HypothesisTest

This class is used through its static methods in order to evaluate some hypothesis on samples : independence or monotonous relation.

##### Some methods :

###### *ChiSquared*

###### Usage :

*HypothesisTest().ChiSquared(firstSample, secondSample)*

*HypothesisTest().ChiSquared(firstSample, secondSample, level)*

###### Arguments :

*firstSample* : a NumericalSample, of dimension 1 : the first sample tested

*secondSample* : a NumericalSample, of dimension 1 : the second sample tested

*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be < 1, equal to 0.95 by default

**Value** : a TestResult, the structure which contains the result of the test

**Details** : the ChiSquared Test is used to check whether two discrete samples are independent

###### *FullPearson*

###### Usage :

*HypothesisTest().FullPearson(firstSample, secondSample)*

*HypothesisTest().FullPearson(firstSample, secondSample, level)*

###### Arguments :

*firstSample* : a NumericalSample, of dimension  $n \geq 1$  : the first sample tested

*secondSample* : a NumericalSample, of dimension 1 : the second sample tested

*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be < 1, equal to 0.95 by default

**Value** : a TestResultCollection, the structure which contains the results of the successive tests.

**Details** : the FullPearson Test is the independence Pearson test between 2 samples : firstSample of dimension n and secondSample of dimension 1. If firstSample[i] is the numerical sample extracted from firstSample (ith coordinate of each point of the numerical sample), FullPearson performs the Independence Pearson test simultaneously on firstSample[i] and secondSample. For all i, it is supposed that the couple (firstSample[i] and secondSample) is issued from a gaussian vector.

###### *FullRegression*

###### Usage :

*HypothesisTest().FullRegression(firstSample, secondSample)*

*HypothesisTest().FullRegression(firstSample, secondSample, level)*

###### Arguments :

*firstSample* : a NumericalSample, of dimension  $n \geq 1$  : the first sample tested

*secondSample* : a NumericalSample, of dimension 1 : the second sample tested

*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be < 1, equal to 0.95 by default.

**Value :** a TestResultCollection, the structure which contains the results of the successive tests.

**Details :** the FullRegression Test is used to check the quality of the regression model between two samples : firstSample of dimension  $n$  and secondSample of dimension 1. If firstSample[i] is the numerical sample extracted from firstSample (ith coordinate of each point of the numerical sample), FullRegression performs the Regression test simultaneously on all firstSample[i] and secondSample. The Regression test tests if the regression model between two scalar numerical samples is significant. It is based on the deviation analysis of the regression. The Fisher distribution is used.

### *FullSpearman*

**Usage :**

*HypothesisTest().FullSpearman(firstSample, secondSample)*  
*HypothesisTest().FullSpearman(firstSample, secondSample, level)*

**Arguments :**

*firstSample* : a NumericalSample, of dimension  $n \geq 1$  : the first sample tested  
*secondSample* : a NumericalSample, of dimension 1 : the second sample tested  
*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be  $< 1$ , equal to 0.95 by default.

**Value :** a TestResultCollection, the structure which contains the results of the successive tests.

**Details :** the FullSpearman Test is used to check the hypothesis of monotonous relation between samples : firstSample of dimension  $n$  and secondSample of dimension 1. If firstSample[i] is the numerical sample extracted from firstSample (ith coordinate of each point of the numerical sample), FullSpearman performs the Independence Spearman test simultaneously on all firstSample[i] and secondSample.

### *PartialPearson*

**Usage :**

*HypothesisTest().PartialPearson(firstSample, secondSample, selection)*  
*HypothesisTest().PartialPearson(firstSample, secondSample, selection, level)*

**Arguments :**

*firstSample* : a NumericalSample, of dimension  $n \geq 1$  : the first sample tested  
*secondSample* : a NumericalSample, of dimension 1 : the second sample tested  
*selection* : a Indices, array of integers selecting the indices of the coordinates of the first sample which will successively be tested with the second sample through the Pearson Test  
*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be  $< 1$ , equal to 0.95 by default.

**Value :** a TestResultCollection, the structure which contains the results of the successive tests.

**Details :** the PartialPearson Test is the independence Pearson test between 2 samples : firstSample of dimension  $n$  and secondSample of dimension 1. If firstSample[i] is the numerical sample extracted from firstSample (ith coordinate of each point of the numerical sample), PartialPearson performs the Independence Pearson test simultaneously on firstSample[i] and secondSample, for  $i$  in the selection. For all  $i$ , it is supposed that the couple (firstSample[i] and secondSample) is issued from a gaussian vector.

*PartialRegression***Usage :**

*HypothesisTest().PartialRegression(firstSample, secondSample, selection)*  
*HypothesisTest().PartialRegression(firstSample, secondSample, selection, level)*

**Arguments :**

*firstSample* : a NumericalSample, of dimension  $n \geq 1$  : the first sample tested  
*secondSample* : a NumericalSample, of dimension 1 : the second sample tested  
*selection* : a Indices, array of integers selecting the indices of the coordinates of the first sample which will successively be tested with the second sample through the Regression Test  
*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be  $< 1$ , equal to 0.95 by default.

**Value** : a TestResult, the structure which contains the result of the test.

**Details** : the PartialRegression Test is used to check the quality of the regression model between two samples : firstSample of dimension n and secondSample of dimension 1. If firstSample[i] is the numerical sample extracted from firstSample (ith coordinate of each point of the numerical sample), PartialRegression performs the Regression test simultaneously on all firstSample[i] and secondSample, for i in the selection. The Regression test tests if the regression model between two scalar numerical samples is significant. It is based on the deviation analysis of the regression. The Fisher distribution is used.

*PartialSpearman***Usage :**

*HypothesisTest().PartialSpearman(firstSample, secondSample, selection)*  
*HypothesisTest().PartialSpearman(firstSample, secondSample, selection, level)*

**Arguments :**

*firstSample* : a NumericalSample, of dimension  $n \geq 1$  : the first sample tested  
*secondSample* : a NumericalSample, of dimension 1 : the second sample tested  
*selection* : a Indices, array of integers selecting the indices of the coordinates of the first sample which will successively be tested with the second sample through the Spearman Test  
*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be  $< 1$ , equal to 0.95 by default.

**Value** : a TestResultCollection, the structure which contains the results of the successive tests.

**Details** : the PartialSpearman Test is used to check the hypothesis of monotonous relation between samples : firstSample of dimension n and secondSample of dimension 1. If firstSample[i] is the numerical sample extracted from firstSample (ith coordinate of each point of the numerical sample), PartialSpearman performs the Independence Spearman test simultaneously on firstSample[i] and secondSample, for i in the selection.

*Pearson***Usage :**

*HypothesisTest().Pearson(firstSample, secondSample)*  
*HypothesisTest().Pearson(firstSample, secondSample, level)*

**Arguments :**

*firstSample* : a NumericalSample, of dimension 1 : the first sample tested

*secondSample* : a NumericalSample, of dimension 1 : the second sample tested

*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be < 1, equal to 0.95 by default.

**Value** : a TestResult, the structure which contains the result of the test.

**Details** : the Test is used to check whether two samples which form a gaussian vector are independent (based on the evaluation of the linear correlation coefficient).

### *Smirnov*

**Usage** :

*HypothesisTest().Smirnov(firstSample, secondSample)*

*HypothesisTest().Smirnov(firstSample, secondSample, level)*

**Arguments** :

*firstSample* : a NumericalSample, of dimension 1 : the first sample tested

*secondSample* : a NumericalSample, of dimension 1 : the second sample tested

*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be < 1, equal to 0.95 by default.

**Value** : a TestResult, the structure which contains the result of the test.

**Details** : the Smirnov Test is used to check whether two continuous scalar samples (of sizes not necessarily equal) follow the same distribution.

### *Spearman*

**Usage** :

*HypothesisTest().Spearman(firstSample, secondSample)*

*HypothesisTest().Spearman(firstSample, secondSample, level)*

**Arguments** :

*firstSample* : a NumericalSample, of dimension 1 : the first sample tested

*secondSample* : a NumericalSample, of dimension 1 : the second sample tested

*level* : a positive real value, the threshold p-value of the test ( = 1- first type risk), must be < 1, equal to 0.95 by default.

**Value** : a TestResult, the structure which contains the result of the test.

**Details** : the Spearman Test is used to check whether two scalar samples have a monotonous relation.

## 4.5 Linear model

### 4.5.1 LinearModelFactory

This class is used in order to create a linear model from numerical samples.

**Usage :** *LinearModelFactory()*

**Arguments :** none

**Some methods :**

*buildLM*

**Usage :**

*buildLM(sampleX, sampleY)*

*buildLM(sampleX, sampleY, level)*

**Arguments :**

*sampleX* : a NumericalSample, of dimension  $n \geq 1$

*sampleY* : a NumericalSample, of dimension 1

*level* : the level value of the confidence intervals of each coefficient of the linear model, equal to 0.95 by default

**Value :** a LinearModel, the linear model built from the samples (*sampleX, sampleY*) :  $Y = a_0 + \sum_i a_i X_i + \varepsilon$ , where  $\varepsilon$  is the aleatory residual with zero mean.



#### 4.5.2 LinearModel

**Usage :** A LinearModel is created through the method *buildLM* of a LinearModelFactory.

**Some methods :**

*getConfidenceIntervals*

**Usage :** *getConfidenceIntervals()*

**Arguments :** none

**Value :** a ConfidenceIntervalPersistentCollection, the collection of the confidence intervals of the linear model coefficients, corresponding to the level precised when the LinearModel class has been created through the method *buildLM*

*getPValues*

**Usage :** *getPValues()*

**Arguments :** none

**Value :** a NumericalScalarPersistentCollection, the collection of the p-values of the linear model coefficients

*getPredict*

**Usage :** *getPredict(sampleX)*

**Arguments :** *sampleX* : a NumericalSample, the sample we want to evaluate the response *Y* on

**Value :** a NumericalSample, of dimension 1, the response *Y* evaluated through the linear model on the sample *sampleX*

*getRegression*

**Usage :** *getRegression()*

**Arguments :** none

**Value :** a NumericalPoint, the coefficients of the linear model :  $(a_0, a_1, \dots, a_n)$

*getResidual*

**Usage :** *getResidual(sampleX, sampleY)*

**Arguments :**

*sampleX* : a NumericalSample, the *sampleX* on which the linear model has been built

*sampleY* : a NumericalSample, the *sampleY* on which the linear model has been built

**Value :** a NumericalPoint, the residuals

## 5 Threshold exceedance probability evaluation with reliability algorithms

### 5.1 Optimisation

#### 5.1.1 NearestPointAlgorithm

**Usage :**

*NearestPointAlgorithm(levelFunction)*

*NearestPointAlgorithm(nearestPointAlgorithmImplementation)*

**Arguments :**

*levelFunction* : a NumericalMathFunction, the constraint function of the constrained optimisation problem

*nearestPointAlgorithmImplementation* : a NearestPointAlgorithmImplementation, the implementation of the nearest point algorithm, which is *Cobyla* , *AbdoRackwitz* or *SQP*.

**Some methods :**

*getLevelFunction*

**Usage :** *getLevelFunction()*

**Arguments :** none

**Value :** a NumericalMathFunction, the constraint function of the constrained optimisation problem

*getLevelValue*

**Usage :** *getLevelFunction()*

**Arguments :** none

**Value :** a real value, the level value of the constraint function in the constrained optimisation problem

*getMaximumAbsoluteError*

**Usage :** *getMaximumAbsoluteError()*

**Arguments :** none

**Value :** a positive real value, the maximum absolute error : maximum distance between two successive iterates

*getMaximumConstraintError*

**Usage :** *getMaximumConstraintError()*

**Arguments :** none

**Value :** a positive real value, the maximum absolute value of the constraint function minus the level value

*getMaximumIterationsNumber*

**Usage :** *getMaximumIterationsNumber()*

**Arguments :** none

**Value :** an integer, the maximum number of iterations of the algorithm

*getMaximumRelativeError*

**Usage :** *getMaximumRelativeError()*

**Arguments :** none

**Value :** a real value, the maximum relative distance between two successive iterates (with regards the second iterate)

*getMaximumResidualError*

**Usage :** *getMaximumResidualError()*

**Arguments :** none

**Value :** a real value, the maximum orthogonality error (lack of orthogonality between the vector Center - Iterate and the constraint surface)

*getResult*

**Usage :** *getResult()*

**Arguments :** none

**Value :** a *NearestPointAlgorithmImplementationResult*, the structure containing all the results of the constrained optimisation problem

*getStartingPoint*

**Usage :** *getStartingPoint()*

**Arguments :** none

**Value :** a *NumericalPoint*, the starting point of the constrained optimisation research

*run*

**Usage :** *run()*

**Arguments :** none

**Value :** none

**Role :** it creates a *NearestPointAlgorithmImplementationResult*, the optimisation result wich is accesible with the method *getResult()*.

Each *getMethod* is associated to a *setMethod*.

### 5.1.2 Cobyala

**Usage :**

*Cobyala()*

*Cobyala(specificParameters, levelFunction)*

**Arguments :**

*specificParameters* : a CobyalaSpecificParameters, the list of the parameters specific to the Cobyala algorithm

*levelFunction* : a NumericalMathFunction, the constraint function of the constrained optimisation problem

**Details :** When no argument is specified, the parameters will have to be fulfilled after, for example, when used in a FORM algorithm (see the corresponding Use Case).

**Some methods :**

*getSpecificParameters*

**Usage :** *getSpecificParameters()*

**Arguments :** none

**Value :** a CobyalaSpecificParameters, the list of the parameters specific to the Cobyala algorithm

This *getMethod* is associated to a *setMethod*.

### 5.1.3 CobySpecificParameters

Usage :

*CobySpecificParameters()*

*CobySpecificParameters(rhoBeg)*

**Arguments :** *rhoBeg* : a real positive strictly value, a reasonable initial step to the variables. When not fulfilled, by default equal to 0.1

**Some methods :**

*getRhoBeg*

**Usage :** *getRhoBeg()*

**Arguments :** none

**Value :** a real value  $>0$ , a reasonable initial step to the variables

This *getMethod* is associated to a *setMethod*.

#### 5.1.4 AbdoRackwitz

**Usage :**

*AbdoRackwitz()*  
*AbdoRackwitz(specificParameters, levelFunction)*

**Arguments :**

*specificParameters* : a *AbdoRackwitzSpecificParameters*, the list of the parameters specific to the AbdoRackwitz algorithm  
*levelFunction* : a *NumericalMathFunction*, the constraint function of the constrained optimisation problem

**Details :** When no argument is specified, the parameters will have to be fulfilled after, for example, when used in a FORM algorithm (see the corresponding Use Case).

The AbdoRackwitz algorithm is a gradient-based constrained optimization method, thus the *NumericalMathFunction* has to provide its gradient:

- If the function is a meta-model generated by OpenTURNS, the analytical gradient is automatically provided.
- If it is an analytical function, OpenTURNS provides a gradient based on the centered finite difference method, that the user can change to better fit its will by another *GradientImplementation* (e.g. constructed by another finite difference method).
- If the function is loaded thanks to a wrapper and if the wrapper provide an implementation of the gradient, OpenTURNS uses it. If there is no gradient provided, OpenTURNS provides a gradient based on centered finite difference method, but with a parameterization different from the case of analytical functions (assuming a lower precision for the function evaluation than in the analytical case).

Be aware of the potential pitfalls associated with the use of finite differences and check the value of the finite difference epsilon for each dimension if AbdoRackwitz algorithm fails to converge.

**Some methods :**

*getSpecificParameters*

**Usage :** *getSpecificParameters()*

**Arguments :** none

**Value :** a *AbdoRackwitzSpecificParameters*, list of the parameters specific to the AbdoRackwitz algorithm

This *getMethod* is associated to a *setMethod*.

### 5.1.5 AbdoRackwitzSpecificParameters

Usage :

*AbdoRackwitzSpecificParameters()*

*AbdoRackwitzSpecificParameters(Omega, Smooth, Tau)*

Arguments :

*Omega* : a real strictly positive value, the Armijo factor. It must be  $< 1$  and should be rather small.  
When not fulfilled, by default equal to  $10^{-4}$

*Smooth* : a real value  $> 1$ , the increasing rate of the penalisation coefficient for the line search. It must be  $> 1$  and should be rather near 1. When not fulfilled, by default equal to 1.2

*Tau* : a real positive value, the multiplicative decrease of the linear step. It must be  $< 1$ . When not fulfilled, by default equal to 0.5

Some methods :

*getOmega*

**Usage :** *getOmega()*

**Arguments :** none

**Value :** a real value between 0 and 1, the Armijo factor

*getSmooth*

**Usage :** *getSmooth()*

**Arguments :** none

**Value :** a real value  $> 1$ , the increasing rate of the penalisation coefficient for the line search

*getTau*

**Usage :** *getTau()*

**Arguments :** none

**Value :** a real value between 0 and 1, the multiplicative decrease of the linear step

This *getMethod* is associated to a *setMethod*.

### 5.1.6 SQP

**Usage :** *SQP(specificParameters,levelFunction)*

**Arguments :**

*specificParameters* : a SQPSpecificParameters, list of the parameters specific to the SQP algorithm

*levelFunction* : a NumericalMathFunction, the constraint function of the constrained optimisation problem

**Some methods :**

*getSpecificParameters*

**Usage :** *getSpecificParameters()*

**Arguments :** none

**Value :** a SQPSpecificParameters, list of the parameters specific to the SQP algorithm

This *getMethod* is associated to a *setMethod*.



### 5.1.7 SQPSpecificParameters

**Usage :**

*SQPSpecificParameters()*

*SQPSpecificParameters(Omega, Smooth, Tau)*

**Arguments :**

*Omega* : a real value by default equal to  $10^{-4}$ , Armijo factor, must be  $> 0$  and  $< 1$  but rather small

*Smooth* : a real value by default equal to 1.2, the increasing rate of the penalisation coefficient for the line search, must be  $> 1$  but rather near 1

*Tau* : a real value by default equal to 0.5, multiplicative decrease of the linear step, must be  $> 0$  and  $< 1$

**Some methods :**

*getOmega*

**Usage :** *getOmega()*

**Arguments :** none

**Value :** a real value between 0 and 1, Armijo factor

*getSmooth*

**Usage :** *getSmooth()*

**Arguments :** none

**Value :** a real value  $> 1$ , the increasing rate of the penalisation coefficient for the line search

*getTau*

**Usage :** *getTau()*

**Arguments :** none

**Value :** a real value between 0 and 1, multiplicative decrease of the linear step

This *getMethod* is associated to a *setMethod*.

### 5.1.8 NearestPointAlgorithmImplementationResult

**Usage :** structure created by the method `run()` of a `NearestPointAlgorithm` and obtained thanks to the method `getResult()`

**Some methods :**

*getAbsoluteError*

**Usage :** `getAbsoluteError()`

**Arguments :** none

**Value :** a `NumericalScalar`, the absolute error : the distance between the two last successive iterates when the algorithm stops

*getConstraintError*

**Usage :** `getConstraintError()`

**Arguments :** none

**Value :** a real value, the absolute value of the constraint function minus the level value at the last iterate point when the algorithm stops

*getIterationsNumber*

**Usage :** `getIterationsNumber()`

**Arguments :** none

**Value :** an integer, the number of performed iterations of the algorithm when the algorithm stops

*getMinimizer*

**Usage :** `getMinimizer()`

**Arguments :** none

**Value :** a `NumericalPoint`, the last iterate when the algorithm stops (the solution of the optimisation problem)

*getRelativeError*

**Usage :** `getRelativeError()`

**Arguments :** none

**Value :** a real value, the relative distance between the two last successive iterates (with regards the last iterate)

*getResidualError*

**Usage :** `getResidualError()`

**Arguments :** none

**Value :** a real value, the orthogonality error of the solution point (lack of orthogonality between the vector Center - Last Iterate and the constraint surface)

## 5.2 Reliability Algorithms

### 5.2.1 Analytical

**Usage :** *Analytical(nearestPointAlgorithm, event, physicalStartingPoint)*

**Arguments :**

*nearestPointAlgorithm* : a NearestPointAlgorithm, the optimisation algorithm which will be used to research the design point

*event* : a Event, the event we want to evaluate the probability

*physicalStartingPoint* : a NumericalPoint, the starting point of the optimisation research, declared in the physical space

**Some methods :**

*getAnalyticalResult*

**Usage :** *getAnalyticalResult()*

**Arguments :** none

**Value :** a AlgorithmAnalyticalResult, the result structure which contains results

*getEvent*

**Usage :** *getEvent()*

**Arguments :** none

**Value :** a Event, the event we want to evaluate the probability

*getNearestPointAlgorithm*

**Usage :** *getNearestPointAlgorithm()*

**Arguments :** none

**Value :** a NearestPointAlgorithm, the optimisation algorithm which will be used to research the design point

*getPhysicalStartingPoint*

**Usage :** *getPhysicalStartingPoint()*

**Arguments :** none

**Value :** a NumericalPoint, the starting point of the optimisation research, declared in the physical space

*run*

**Usage :** *run()*

**Arguments :** none

**Value :** it performs the research design point and creates a AnalyticalResult, the structure result wich is accesible with the method *getAnalyticalResult()*.

The methods *getEvent*, *getNearestPointAlgorithm* and *getPhysicalStartingPoint* are associated to a *setMethod*.

**Derivative Classes :** FORM ans SORM

### 5.2.2 AnalyticalResult

**Usage :** structure created by the method `run()` of a `Analytical` and obtained thanks to the method `getAnalyticalResult()`

**Some methods :**

*drawHasoferReliabilityIndexSensitivity*

**Usage :** *drawHasoferReliabilityIndexSensitivity()*

**Arguments :** none

**Value :** a `GraphCollection` (the collection of two barplots) drawing the sensitivity of the Hasofer Reliability Index to the parameters of the marginals of the probabilistic input vector (first graph) and to the parameters of the dependence structure of the probabilistic input vector (second graph).

*drawImportanceFactors*

**Usage :** *drawImportanceFactors()*

**Arguments :** none

**Value :** a `Graph`, the pie of the importance factors of the probabilistic variables

*getHasoferReliabilityIndex*

**Usage :** *getHasoferReliabilityIndex()*

**Arguments :** none

**Value :** a real positive value, the Hasofer Reliability Index

*getHasoferReliabilityIndexSensitivity*

**Usage :** *getHasoferReliabilityIndexSensitivity()*

**Arguments :** none

**Value :** a `Sensitivity`, the sensitivities of the Hasofer Reliability Index to the parameters of the probabilistic input vector (marginals and dependence structure)

*getImportanceFactors*

**Usage :** *getImportanceFactors()*

**Arguments :** none

**Value :** a `NumericalPoint`, the importance factors of probabilistic variables

*getIsStandardPointOriginInFailureSpace*

**Usage :** *getIsStandardPointOriginInFailureSpace()*

**Arguments :** none

**Value :** a boolean which indicates whether the origine of the standard space is in the failure space

*getLimitStateVariable*

**Usage :** *getLimitStateVariable()*

**Arguments :** none

**Value :** a Event, the event we evaluated the probability

*getPhysicalSpaceDesignPoint*

**Usage :** *getPhysicalSpaceDesignPoint()*

**Arguments :** none

**Value :** a NumericalPoint, the starting point of the optimisation research, declared in the physical space

*getStandardSpaceDesignPoint*

**Usage :** *getStandardSpaceDesignPoint()*

**Arguments :** none

**Value :** a NumericalPoint, the starting point of the optimisation research, declared in the standard space

**Derivative Classes :** FORMResult and SORMResult

### 5.2.3 Event

**Usage :**

*Event(antecedent, comparisonOperator, threshold)*

*Event(antecedent, comparisonOperator, threshold, name)*

**Arguments :**

*antecedent* : a RandomVector, of dimension 1 : the output variable of interest

*comparisonOperator* : a ComparisonOperator, the comparison operator which is equal to *Less*, *Greater*, *LessOrEqual* or *GreaterOrEqual*

*threshold* : a real value, the threshold we want to compare to *antecedent*

*name* : a string, the name of the event

**Some methods :**

*getDimension*

**Usage :** *getDimension()*

**Arguments :** none

**Value :** an integer, the dimension of the probabilistic input vector

*getNumericalSample*

**Usage :** *getNumericalSample(size)*

**Arguments :** *size* : an integer, the size of the numerical sample generated

**Value :** a NumericalSample filled with boolean values (1 for the realisation of the event and 0 else)  
: *size* realisations of the event (considered as a Bernoulli variable)

*getOperator*

**Usage :** *getOperator()*

**Arguments :** none

**Value :** a ComparisonOperator, the comparison operator of the event

*getRealization*

**Usage :** *getRealization()*

**Arguments :** none

**Value :** a NumericalPoint of dimension 1, filled with a boolean value (1 for the realisation of the event and 0 else) : one realisation of the event (considered as a Bernoulli variable)

*getThreshold*

**Usage :** *getThreshold()*

**Arguments :** none

**Value :** a real value, the threshold of the event

**Derivative Class :** StandardEvent

### 5.2.4 StandardEvent

This class inherits from Event.

#### Usage :

```
StandardEvent(antecedent, comparisonOperator, threshold)  
StandardEvent(antecedent, comparisonOperator, threshold, name)  
StandardEvent(event)
```

#### Arguments :

*antecedent* : a RandomVector, of dimension 1 : the output variable of interest  
*comparisonOperator* : a ComparisonOperator, the comparison operator which is equal to *Less*, *Greater*, *LessOrEqual* or *GreaterOrEqual*  
*threshold* : a real value, the threshold we want to compare to *antecedent*  
*name* : a string, the name of the event  
*event* : a Event, the physical event associated to the standard event

### 5.2.5 FORM

This class inherits from Analytical.

**Usage :** *FORM(nearestPointAlgorithm, event, physicalStartingPoint)*

**Arguments :**

*nearestPointAlgorithm* : a NearestPointAlgorithm, the optimisation algorithm which will be used to research the design point

*event* : a Event, the event in the physical space we want to evaluate the probability

*physicalStartingPoint* : a NumericalPoint, the starting point of the optimisation research, declared in the physical space

**Some methods :**

*getResult*

**Usage :** *getResult()*

**Arguments :** none

**Value :** a FORMResult, structure containing all the results of the FORM analysis

*run*

**Usage :** *run()*

**Arguments :** none

**Value :** it creates a FORMResult, the optimisation result wich is accesible with the method getResult().



### 5.2.6 FORMResult

This class inherits from AnalyticalResult.

**Usage :** structure created by the method `run()` of a FORM and obtained thanks to the method `getResult()`

**Some methods :**

*drawEventProbabilitySensitivity*

**Usage :** *drawEventProbabilitySensitivity()*

**Arguments :** none

**Value :** a GraphCollection (the collection of two barplots) drawing the sensitivities of the FORM Probability with regards to the parameters of the probabilistic input vector (first graph) and to parameters of the dependence structure of the probabilistic input vector (second graph)

*getEventProbability*

**Usage :** *getEventProbability()*

**Arguments :** none

**Value :** a positive real value, the FORM probability of the event

*getEventProbabilitySensitivity*

**Usage :** *getEventProbabilitySensitivity()*

**Arguments :** none

**Value :** a Sensitivity, the sensitivities of the FORM Probability with regards to the parameters of the probabilistic input vector and to parameters of the dependence structure of the probabilistic input vector

*getGeneralisedReliabilityIndex*

**Usage :** *getGeneralisedReliabilityIndexHohenBichler()*

**Arguments :** none

**Value :** a real value, the generalised reliability index evaluated from the FORM Probability. The generalised reliability index from the FORM probability is equal to  $\pm$  the Hasofer reliability index according to the fact the standard space center fulfills the event or not

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### 5.2.7 SORM

This class inherits from Analytical.

**Usage :** *SORM(nearestPointAlgorithm, event, physicalStartingPoint)*

**Arguments :**

*nearestPointAlgorithm* : a NearestPointAlgorithm, the optimisation algorithm which will be used to research the design point

*event* : a Event, the event in the physical space we want to evaluate the probability

*physicalStartingPoint* : a NumericalPoint, the starting point of the optimisation research, declared in the physical space

**Some methods :**

*getResult*

**Usage :** *getResult()*

**Arguments :** none

**Value :** a SORMResult, structure containing all the results of the SORM analysis

*run*

**Usage :** *run()*

**Arguments :** none

**Value :** it creates a SORMResult, the optimisation result wich is accesible with the method getResult().

### 5.2.8 SORMResult

This class inherits from AnalyticalResult.

**Usage :** structure created by the method `run()` of a SORM and obtained thanks to the method `getResult()`

**Some methods :**

*getEventProbabilityBreitung*

**Usage :** `getEventProbabilityBreitung()`

**Arguments :** none

**Value :** a positive real value, the SORM Probability according to the Breitung approximation

*getEventProbabilityHohenBichler*

**Usage :** `getEventProbabilityHohenBichler()`

**Arguments :** none

**Value :** a positive real value, the SORM Probability according to the Hohen Bichler approximation

*getEventProbabilityTvedt*

**Usage :** `getEventProbabilityTvedt()`

**Arguments :** none

**Value :** a positive real value, the SORM Probability according to the Tvedt approximation

*getGeneralisedReliabilityIndexBreitung*

**Usage :** `getGeneralisedReliabilityIndexBreitung()`

**Arguments :** none

**Value :** a real value, the generalised reliability index evaluated from the Breitung SORM Probability (positive or negative according to the fact the standard space center fulfills the event or not)

*getGeneralisedReliabilityIndexHohenBichler*

**Usage :** `getGeneralisedReliabilityIndexHohenBichler()`

**Arguments :** none

**Value :** a real value, the generalised reliability index evaluated from the Hohen Bichler SORM Probability (positive or negative according to the fact the standard space center fulfills the event or not)

*getGeneralisedReliabilityIndexTvedt*

**Usage :** `getGeneralisedReliabilityIndexTvedt()`

**Arguments :** none

**Value :** a real value, the generalised reliability index evaluated from the Tvedt SORM Probability (positive or negative according to the fact the standard space center fulfills the event or not)

## 5.3 The Strong Max Test

### 5.3.1 StrongMaxTest

Usage :

```
StrongMaximumTest(event, standardSpaceDesignPoint, importanceLevel, ...
                  accuracyLevel, confidenceLevel)
StrongMaximumTest(event, standardSpaceDesignPoint, importanceLevel, ...
                  accuracyLevel, pointNumber)
```

Arguments :

*event* : a StandardEvent,  
*standardSpaceDesignPoint* : a NumericalPoint,  
*importanceLevel* : a positive real value, the importance level  $\varepsilon$ .  
*accuracyLevel* : a positive real value, the accuracy Level  $\tau$ . It is recommended to take  $\tau \leq 4$ .  
*confidenceLevel* : a positive real value, the confidenceLevel  $(1 - q)$ , must be  $< 1$ .  
*pointNumber* : the number of points used to perform the Strong Maximum Test on which the limit state function is evaluated

Some methods :

*getAccuracyLevel*

**Usage :** *getAccuracyLevel()*

**Arguments :** none

**Value :** a positive real value, the accuracy Level  $\tau$

*getConfidenceLevel*

**Usage :** *getConfidenceLevel()*

**Arguments :** none

**Value :** a positive real value, the confidenceLevel  $(1 - q)$

*getEvent*

**Usage :** *getEvent()*

**Arguments :** none

**Value :** a StandardEvent, the event in the standard space on which is based the Strong Maximum Test

*getFarDesignPointVerifyingEventPoints*

**Usage :** *getFarDesignPointVerifyingEventPoints()*

**Arguments :** none

**Value :** a NumericalSample, the list of points of the discretised sphere which are out of the vicinity of the standard design point and which verify the event

*getFarDesignPointVerifyingEventValues*

**Usage :** *getFarDesignPointVerifyingEventValues()*

**Arguments :** none

**Value :** a NumericalSample, the list of the values of the limit state function on the points of the discretised sphere which are out of the vicinity of the standard design point and which verify the event

*getFarDesignPointViolatingEventPoints*

**Usage :** *getFarDesignPointViolatingEventPoints()*

**Arguments :** none

**Value :** a NumericalSample, the list of points of the discretised sphere which are out of the vicinity of the standard design point and which don't verify the event

*getFarDesignPointViolatingEventValues*

**Usage :** *getFarDesignPointViolatingEventValues()*

**Arguments :** none

**Value :** a NumericalSample, the list of the values of the limit state function on the points of the discretised sphere which are out of the vicinity of the standard design point and which don't verify the event

*getImportanceLevel*

**Usage :** *getImportanceLevel()*

**Arguments :** none

**Value :** a positive real value, the importance level  $\varepsilon$

*getNearDesignPointVerifyingEventPoints*

**Usage :** *getNearDesignPointVerifyingEventPoints()*

**Arguments :** none

**Value :** a NumericalSample, the list of points of the discretised sphere which are inside the vicinity of the standard design point and which verify the event

*getNearDesignPointVerifyingEventValues*

**Usage :** *getNearDesignPointVerifyingEventValues()*

**Arguments :** none

**Value :** a NumericalSample, the list of the values of the limit state function on the points of the discretised sphere which are inside the vicinity of the standard design point and which verify the event

*getNearDesignPointViolatingEventPoints*

**Usage :** *getNearDesignPointViolatingEventPoints()*

**Arguments :** none

**Value :** a NumericalSample, the list of points of the discretised sphere which are out of the vicinity of the standard design point and which don't verify the event

*getNearDesignPointViolatingEventValues*

**Usage :** *getNearDesignPointViolatingEventValues()*

**Arguments :** none

**Value :**

a NumericalSample, the list of the values of the limit state function on the points of the discretised sphere which are inside the vicinity of the standard design point and which don't verify the event

*getPointNumber*

**Usage :** *getPointNumber()*

**Arguments :** none

**Value :** an integer, the number of points used to perform the Strong Maximum Test, evaluated by the limit state function

*getStandardSpaceDesignPoint*

**Usage :** *getStandardSpaceDesignPoint()*

**Arguments :** none

**Value :** a NumericalPoint, the standard space design point

*run*

**Usage :** *run()*

**Arguments :** none

**Value :** it performs the Strong Maximum Test.

## 6 Threshold exceedance probability evaluation with simulation

## 6.1 HistoryStrategy

Four storage strategies are proposed by Open TURNS, in order to store the numerical sample (in and out) used to evaluate the probability estimator, and to store the values of the estimator and its standard deviation used to draw the convergence graph.

**Usage :**

```
HistoryStrategy(Null())
HistoryStrategy(Full())
HistoryStrategy(Compact(Ncompact))
HistoryStrategy>Last(Nlast))
```

**Arguments :**

*Ncompact* : an integer, the number of points stored is in  $[Ncompact, 2 * Ncompact]$ . If the sample size is smaller than *Ncompact*, the entire sample is stored.

*Nlast* : an integer, the number of points stored. If the sample size is smaller than *Nlast*, the entire sample is stored.

**Details :**

*HistoryStrategy(Null())* : nothing is stored.

*HistoryStrategy(Full())* : each point is stored. Be careful! The memory will be exhausted for huge samples.

*HistoryStrategy(Compact(Ncompact))* : a regularly spaced sub-sample is stored, with a maximum size if the sample is large enough.

*HistoryStrategy>Last(Nlast))* : only the *Nlast* last points are stored.

**Some methods :**

*getSample*

**Usage :** *getSample()*

**Arguments :** none

**Value :** a NumericalSample containing the object stored (the in and out numerical sample or the values of the probability estimator and its standard deviation)

## 6.2 RandomGenerator

**Usage :** *RandomGenerator()*

**Arguments :** none

**Some methods :**

*Generate*

**Usage :**

*Generate()*

*Generate(size)*



**Arguments :** *size* : an integer, the number of realisations required. When not fulfilled, by default equal to 1

**Value :** a NumericalPoint, the list of the required realisations of a uniform distribution on  $[0,1]$

#### *GetState*

**Usage :** *GetState()*

**Arguments :** : none

**Value :** a RandomGeneratorState, the state of the random generator

#### *SetSeed*

**Usage :** *SetSeed(n)*

**Arguments :** *n* : an integer which enables an easy initialisation of the random generator

**Value :** none. It initialises the state of the random generator

#### *SetState*

**Usage :** *SetState(state)*

**Arguments :** *state* : a RandomGeneratorState, the state of the random generator

**Value :** none. It initialises the state of the random generator

The *GetState* method is associated to a *SetState* one.

### 6.3 Wilks

This class is a static class which enables the evaluation of the Wilks number : the minimal sample size  $N_{\alpha,\beta,i}$  to perform in order to guarantee that the empirical quantile  $\alpha$ , noted  $\tilde{q}_\alpha(N_{\alpha,\beta,i})$  evaluated with the  $(n-i)$ th maximum of the sample, noted  $X_{N-i}$  be greater than the theoretical quantile  $q_\alpha$  with a probability at least  $\beta$  :

$$Prob(\tilde{q}_\alpha(N_{\alpha,\beta,i}) = X_{n-i} > q_\alpha) > \beta$$

This class proposes one method :

*ComputeSampleSize*

**Usage :** *ComputeSampleSize(alpha, beta, i)*

**Arguments :**

*alpha* : a real value, order of the quantile we want the evaluate, must be  $> 1$  and  $> 0$

*beta* : confidence on the evaluation of the empirical quantile, must be  $> 1$  and  $> 0$

*i* : rank of the maximum which will evaluate the empirical quantile, by default  $i = 0$  (maximum of the sample)

**Value :** an integer, the Wilks number.

## 6.4 Simulation

**Usage :** *NearestPointAlgorithm(levelFunction)*

**Arguments :** *levelFunction* : a NumericalMathFunction, the constraint function of the constrained optimisation problem

**Some methods :**

*getBlockSize*

**Usage :** *getBlockSize()*

**Arguments :** none

**Value :** an integer, the number of terms in the probability simulation estimator grouped together

**Details :** for Monte Carlo, LHS and Importance Sampling methods, we recommend to use BlockSize = number of available CPU ; for the Directional Sampling, we recommend to use BlockSize = 1

*getConvergenceStrategy*

**Usage :** *getConvergenceStrategy()*

**Arguments :** none

**Value :** a HistoryStrategy, the storage strategy used to store the values of the probability estimator and its variance during the simulation algorithm

*getEvent*

**Usage :** *getEvent()*

**Arguments :** none

**Value :** an Event, which we want to evaluate the probability

*getInputStrategy*

**Usage :** *getInputStrategy()*

**Arguments :** none

**Value :** a HistoryStrategy, the storage strategy used to store the input random vector sample and the output random vector sample used to evaluate the probability estimator of the event probability

*getMaximumCoefficientOfVariation*

**Usage :** *getMaximumCoefficientOfVariation()*

**Arguments :** none

**Value :** a real value, the maximum coefficient of variation of the simulated sample

*getMaximumOuterSampling*

**Usage :** *getMaximumOuterSampling()*

**Arguments :** none

**Value :** an integer, the maximum number of groups of terms in the probability simulation estimator

**Details :** for Monte Carlo, LHS and Importance Sampling methods, the maximum number of evaluations of the limit state function defining the event is : MaximumOuterSampling \* BlockSize

*getOutputStrategy***Usage :** *getOutputStrategy()***Arguments :** none**Value :** a HistoryStrategy, the storage strategy used to store the input random vector sample and the output random vector sample used to evaluate the probability estimator of the event probability*getResult***Usage :** *getResult()***Arguments :** none**Value :** a SimulationResult, the structure containing all the results obtained after simulation and created by the method run()*run***Usage :** *run()***Arguments :** none**Value :** it launches the simulation and creates a SimulationResult, structure containing all the results obtained after simulation*setConvergenceStrategy***Usage :** *setConvergenceStrategy(myHistoryStrategy)***Arguments :** *myHistoryStrategy* : a HistoryStrategy, the storage strategy used to store the values of the probability estimator and its variance during the simulation algorithm.**Value :** none*setInputOutputStrategy***Usage :** *setInputOutputStrategy(myHistoryStrategy)***Arguments :** *myHistoryStrategy* : a HistoryStrategy, the storage strategy used to store the input random vector sample and the output random vector sample used to evaluate the probability estimator of the event probability**Value :** none

Only the *getBlockSize*, *getMaximumCoefficientOfVariation*, *getMaximumOuterSampling* methods have a *setMethod* associated.

**Derivative Classes :**

MonteCarlo

LHS

DirectionnalSampling

ImportanceSampling

## 6.5 MonteCarlo

This class inherits from Simulation.

**Usage :** *MonteCarlo(event)*

**Arguments :** *event* : an Event, the event we want to evaluate the probability

## 6.6 LHS

This class inherits from Simulation.

**Usage :**  $LHS(event)$

**Arguments :** *event* : an Event, the event we want to evaluate the probability

## 6.7 DirectionalSampling

### 6.7.1 DirectionalSampling

This class inherits from Simulation.

The Directional Sampling simulation operates in the standard space.

#### Usage :

*DirectionalSampling(event)*

*DirectionalSampling(event, rootStrategy, samplingStrategy)*

#### Arguments :

*event* : an Event, the event we want to evaluate the probability

*rootStrategy* : a RootStrategy, the strategy adopted to evaluate the intersections of each direction with the limit state function and take into account the contribution of the direction to the event probability. By default, *rootStrategy* = *RootStrategy(SafeAndSlow)*

*samplingStrategy* : a SamplingStrategy, the strategy adopted to sample directions. By default, *samplingStrategy* = *SamplingStrategy(RandomDirection)*.

#### Some methods :

*getRootStrategy*

**Usage :** *getRootStrategy()*

**Arguments :** none

**Value :** a RootStrategy, the root strategy adopted

*getSamplingStrategy*

**Usage :** *getSamplingStrategy()*

**Arguments :** none

**Value :** a SamplingStrategy, the direction sampling strategy adopted

Each *getMethod* is associated to a *setMethod*.

### 6.7.2 RootStrategy

Usage :

*RootStrategy()*

*RootStrategy(rootStrategyImplementation)*

Arguments :

*rootStrategyImplementation* : a *RootStrategyImplementation*, the implementation of the root strategy adopted, which is *RiskyAndFast*, *MediumSafe* or *SafeAndSlow*

When not fulfilled, by default, *rootStrategyImplementation* = *SafeAndSlow*.

Some methods :

*getMaximumDistance*

Usage : *getMaximumDistance()*

Arguments : none

Value : a positive real value, the distance from the center of the standard space until which we research an intersection with the limit state function along each direction. By default, the maximum distance is equal to 8

*getOriginValue*

Usage : *getOriginValue()*

Arguments : none

Value : a real value, the value of the limit state function at the center of the standard space

*getStepSize*

Usage : *getStepSize()*

Arguments : none

Value : a real value, the length of each segment inside which the root research is performed.

Each *getMethod* is associated to a *setMethod*.



### 6.7.3 RiskyAndFast

The RiskyAndFast strategy is the following : for each direction, we check whether there is a sign changement of the standard limit state function between the maximum distant point (at distance *MaximumDistance* from the center of the standard space) and the center of the standard space.

In case of sign changement, we research one root in the segment [origine, maximum distant point] with the selectionned non linear solver.

As soon as founded, the segment [root, infinity point] is considered within the failure space.

It inherits from the methods of the RootStrategy class.

#### Usage :

*RiskyAndFast()*

*RiskyAndFast(solver)*

*RiskyAndFast(solver, maximumDistance, stepSize)*

#### Arguments :

*solver* : a Solver, the non linear solver used to research the intersection of the limit state function with the direction, on each segment of length *stepSize*, between the center of the space and *maximumDistance* (root research)

*maximumDistance* : a real strictly positive value, the maximum distance within wich the root research is perfomed along each direction

*stepSize* : a real value, the length of each segment along a direction inside which the root research is performed

By default, *solver* = *Brent*, *maximumDistance* = 8, *stepSize* = 1

#### Some methods :

*getSolver*

**Usage :** *getSolver()*

**Arguments :** none

**Value :** a Solver, the non linear solver wich will research the root in a segment

*solve*

**Usage :** *solve(function, value)*

**Arguments :**

*function* : a NumericalMathFunction, from  $\mathbb{R}$  into  $\mathbb{R}$

*value* : a real value

**Value :** a ScalarCollection of dimension 1, the real value  $x$  such as  $function(x) = value$  researched within [*origine*, *maximumDistance*]

Each *getMethod* is associated to a *setMethod*.

#### 6.7.4 MediumSafe

The MediumSafe strategy is the following : for each direction, we go along the direction by step of length *stepSize* from the origin to the maximum distant point (at distance *MaximumDistance* from the center of the standard space) and we check whether there is a sign changement on each segment so formed.

At the first sign changement, we research one root in the concerned segment with the selectionned non linear solver. Then, the segment [root, maximum distant point] is considered within the failure space.

If *stepSize* is small enough, this strategy guarantees us to find the root which is the nearest from the origine.

It inherits from the methods of the RootStrategy class.

#### Usage :

*MediumSafe()*

*MediumSafe(solver)*

*MediumSafe(solver, maximumDistance, stepSize)*

#### Arguments :

*solver* : a Solver, the non linear solver used to research the intersection of the limit state function with the direction, on each segment of length *stepSize*, between the center of the space and *maximumDistance* (root research),

*maximumDistance* : a real strictly positive value, the maximum distance within wich the root research is performed along each direction

*stepSize* : a real value. CARE : this value is not taken into account in the root research : *stepSize* = *maximumDistance* automatically on the algorithm according to this root strategy

By default, *solver* = *Brent*, *maximumDistance* = 8

#### Some methods :

*getSolver*

**Usage :** *getSolver()*

**Arguments :** none

**Value :** a Solver, the non linear solver wich will research the root in a segment

*solve*

**Usage :** *solve(function, value)*

**Arguments :**

*function* : a NumericalMathFunction, from  $\mathbb{R}$  into  $\mathbb{R}$

*value* : a real value

**Value :** a ScalarCollection of dimension 1 (one root) : the real value  $x$  such as  $function(x) = value$  researched the first segment of length *stepsize*, within [*origine*, *maximumDistance*] where a sign changement of *function* has been detected

Each *getMethod* is associated to a *setMethod*.

### 6.7.5 SafeAndSlow

The SafeAndSlow strategy is the following : for each direction, we go along the direction by step of length *stepSize* from the origine to the maximum distant point(at distance *MaximumDistance* from the center of the standard space) and we check whether there is a sign changement on each segment so formed.

We go until the maximum distant point. Then, for all the segments where we detected a the presence of a root, we research the root with the selectionned non linear solver. We evaluate the contribution to the failure probability of each segment.

If *stepSize* is small enough, this strategy guarantees us to find all the roots in the direction and the contribution of this direction to the failure probability is precisely evaluated.

It inherits from the methods of the RootStrategy class.

#### Usage :

```
SafeAndSlow()
SafeAndSlow(solver)
SafeAndSlow(solver, maximumDistance, stepSize)
```

#### Arguments :

*solver* : a Solver, the non linear solver used to research the intersection of the limit state function with the direction, on each segment of length *stepSize*, between the center of the space and *maximumDistance* (root research),

*maximumDistance* : a real strictly positive value, the maximum distance within wich the root research is perfomed along each direction

*stepSize* : a real value, the length of each segment along a direction inside which the root research is performed.

By default, *solver* = *Brent*, *maximumDistance* = 8, *stepSize* = 1

#### Some methods :

*getSolver*

**Usage :** *getSolver()*

**Arguments :** none

**Value :** a Solver, the non linear solver wich will research the root in a segment

*solve*

**Usage :** *solve(function, value)*

**Arguments :**

*function* : a NumericalMathFunction, from  $\mathbb{R}$  into  $\mathbb{R}$

*value* : a real value

**Value :** a ScalarCollection, all the real values  $x$  such as  $function(x) = value$  researched in each segment of length *stepsize*, within [*origine*, *maximumDistance*]

Each *getMethod* is associated to a *setMethod*.

### 6.7.6 SamplingStrategy

Usage :

*SamplingStrategy()*  
*SamplingStrategy(samplingStrategyImplementation)*  
*SamplingStrategy(dimension)*

Arguments :

*samplingStrategyImplementation* : a *SamplingStrategyImplementation*, the implementation of the sampling strategy adopted, which is *RandomDirection*, or *OrthogonalDirection*

*dimension* : an integer, the dimension of the standard space

By default, *samplingStrategyImplementation* = *RandomDirection* and *dimension* = 0 but the dimension automatically updated by the calling class

Some methods :

*getDimension*

**Usage :** *getDimension()*

**Arguments :** none

**Value :** an integer, the dimension of the standard space

Each *getMethod* is associated to a *setMethod*.

### 6.7.7 RandomDirection

The RandomDirection strategy is the following : we generate some points on the sphere unity in the standard space according to the uniform distribution and we consider both opposite directions so built.

It inherits from the methods of the SamplingStrategy class.

**Usage :**

*RandomDirection()*  
*RandomDirection(dimension)*

**Arguments :**

*dimension* : an integer, the dimension of the standard space  
By default, *dimension* = 0 but automatically updated by the calling class

**Some methods :**

*generate*

**Usage :** *generate()*  
**Arguments :** none  
**Value :** a NumericalSample of size 2, two opposite random directions generated

*getUniformUnitVectorRealization*

**Usage :** *getUniformUnitVectorRealization(dimension)*  
**Arguments :** *dimension* : an interger, the dimension of the sphere unity (which is the dimension of the standard space)  
**Value :** a NumericalPoint, a realisation of a vector on the sphere unity, according to the uniform distribution

Each *getMethod* is associated to a *setMethod*.

### 6.7.8 OrthogonalDirection

The OrthogonalDirection strategy is the following : this strategy is parametered by  $k \in \mathbb{N}$ . We generate one direct orthonormalised base  $(e_1, \dots, e_n)$  within the set of orthonormalised bases. We consider all the renormalised linear combinations of  $k$  vectors within the  $n$  vectors of the base, where the coefficients of the linear combinations are equal to  $+1, -1$ . There are  $C_n^k 2^k$  new vectors  $v_i$ . We consider each direction defined by each vector  $v_i$ .

If  $k = 1$ , we consider all the axes of the standard space.

It inherits from the methods of the SamplingStrategy class.

#### Usage :

*OrthogonalDirection()*  
*OrthogonalDirection(dimension, size)*

#### Arguments :

*dimension* : an integer, dimension of the standard space  
*size* : an integer, the number of elements in the linear combinations described here above  
 By default, *size* = 1 and *dimension* = 0 but automatically updated by the calling class.

#### Some methods :

*generate*

**Usage :** *generate()*

**Arguments :** none

**Value :** a NumericalSample, a realisation of a random direction according to the algorithm described here above.

*getUniformUnitVectorRealization*

**Usage :** *getUniformUnitVectorRealization(dimension)*

**Arguments :** *dimension* : an ineteger, the dimension of the sphere unity (dimension of the standard space)

**Value :** a NumericalPoint, a realisation of a vector on the sphere unity, according to the uniform distribution

Each *getMethod* is associated to a *setMethod*.

### 6.7.9 Solver

This class enables to solve 1D non linear equations :

$$f(x) = \text{value}, \text{ for } x \in ]\text{infPoint}, \text{supPoint}[$$

if  $f$  is a function from  $\mathbb{R}$  in  $\mathbb{R}$ ,  $(\text{infPoint}, \text{supPoint}) \in \mathbb{R}^2$  and if  $f$  has a sign changement between the two bounds of the interval  $] \text{infPoint}, \text{supPoint}[$ , which means that  $f(\text{infPoint}) * f(\text{supPoint}) < 0$ .

In particular, it is used in the root research of a directional sampling simulation.

#### Usage :

*Solver(solverImplementation)*

*Solver(absoluteError, relativeError, maximumFunctionEvaluation)*

#### Arguments :

*solverImplementation* : a SolverImplementation, the implementation of particular solver which is *Bisection*, *Brent* or *Secant*,

*absoluteError* : a real positive value, absolute error : distance between two successif iterates at the end point

*relativeError* : a real positive value, relative distance between the two last successif iterates (with regards the last iterate)

*maximumFunctionEvaluation* : an integer, the maximum number of evaluations of the function

#### Some methods :

*getAbsoluteError*

**Usage :** *getAbsoluteError()*

**Arguments :** none

**Value :** a real positive value, the absolute error : distance between two successive iterates at the end point

*getMaximumFunctionEvaluation*

**Usage :** *getMaximumFunctionEvaluation()*

**Arguments :** none

**Value :** an integer, the maximum number of evaluations of the function

*getRelativeError*

**Usage :** *getRelativeError()*

**Arguments :** none

**Value :** a real positive value, the relative distance between the two last successive iterates (with regards the last iterate)

### 6.7.10 Bisection

The Bisection solver is a bisection algorithm.

Usage :

*Bisection()*

*Bisection(absoluteError, relativeError, maximumFunctionEvaluation)*

Arguments :

*absoluteError* : a real positive value, absolute error : distance between two successive iterates at the end point

*relativeError* : a real positive value, relative distance between the two last successive iterates (with regards the last iterate)

*maximumFunctionEvaluation* : an integer, the maximum number of evaluations of the function

By default, *absoluteError* =  $1.e - 5$ , *relativeError* =  $1.e - 5$ , *maximumFunctionEvaluation* = 100

Some methods :

*solve*

Usage :

*solve(function, value, infPoint, supPoint)*

*solve(function, value, infPoint, supPoint, infValue, supValue)*

Arguments :

*function* : a NumericalMathFunction, the function of the equation  $function(x) = value$  we want to solve on  $]infPoint, supPoint[$

*value* : a real value, the value of the equation  $function(x) = value$  we want to solve on  $]infPoint, supPoint[$

*infPoint* : a real value, the lower bound of the interval where we want to solve the equation

*supPoint* : a real value, the upper bound of the interval where we want to solve the equation

*infValue*: a real value, the value of *function* on the point *infPoint* :  $function(infPoint)$ , must be of opposite sign of *supValue*

*supValue*: a real value, a real value, the value of *function* on the point *supPoint* :  $function(supPoint)$ , must be of opposite sign of *infValue*

**Value** : a real value, the result of the root research, in  $]infPoint, supPoint[$

**Details** : If the function *f* is continuous, the Bisection solver will converge towards a root of the equation  $function(x) = value$  on  $]infPoint, supPoint[$ . If not, it will converge towards either a root or a discontinuity point of *f* on  $]infPoint, supPoint[$ . Bisection guarantees a convergence.

Bisection may fail.



### 6.7.11 Brent

The Brent solver is a mix of Bisection, Secant and inverse quadratic interpolation.

#### Usage :

*Brent()*

*Brent(absoluteError, relativeError, maximumFunctionEvaluation)*

#### Arguments :

*absoluteError* : a real positive value, the absolute error : distance between two successive iterates at the end point

*relativeError* : a real positive value, the relative distance between the two last successive iterates (with regards the last iterate)

*maximumFunctionEvaluation* : an integer, the maximum number of evaluations of the function

By default, *absoluteError* =  $1.e - 5$ , *relativeError* =  $1.e - 5$ , *maximumFunctionEvaluation* = 100

#### Some methods :

*solve*

##### Usage :

*solve(function, value, infPoint, supPoint)*

*solve(function, value, infPoint, supPoint, infValue, supValue)*

##### Arguments :

*function* : a NumericalMathFunction, the function of the equation  $function(x) = value$  we want to solve on  $]infPoint, supPoint[$

*value* : a real value, the value of the equation  $f(x) = value$  we want to solve on  $]infPoint, supPoint[$

*infPoint* : a real value, the lower bound of the interval where we want to solve the equation

*supPoint* : a real value, the upper bound of the interval where we want to solve the equation

*infValue*: a real value, the value of *function* on the point *infPoint* :  $function(infPoint)$ , must be of opposite sign of *supValue*

*supValue*: a real value, a real value, the value of *function* on the point *supPoint* :  $function(supPoint)$ , must be of opposite sign of *infValue*

**Value** : the result of the root research, in  $]infPoint, supPoint[$

**Details** : If the function *f* is continuous, the Brent solver will converge towards a root of the equation  $function(x) = value$  on  $]infPoint, supPoint[$ . If not, it will converge towards either a root or a discontinuity point of *f* on  $]infPoint, supPoint[$ . Brent guarantees a convergence.

### 6.7.12 Secant

The Secant solver is based on the evaluation of a segment between the two last iterated points.

#### Usage :

*Secant()*

*Secant(absoluteError, relativeError, maximumFunctionEvaluation)*

#### Arguments :

*absoluteError* : a real positive value, absolute error : distance between two successive iterates at the end point

*relativeError* : a real positive value, relative distance between the two last successive iterates (with regards the last iterate)

*maximumFunctionEvaluation* : an integer, the maximum number of evaluations of the function

By default, *absoluteError* =  $1.e - 5$ , *relativeError* =  $1.e - 5$ , *maximumFunctionEvaluation* = 100

#### Some methods :

*solve*

#### Usage :

*solve(function, value, infPoint, supPoint)*

*solve(function, value, infPoint, supPoint, infValue, supValue)*

#### Arguments :

*function* : a NumericalMathFunction, the function of the equation  $function(x) = value$  we want to solve on  $]infPoint, supPoint[$

*value* : a real value, the value of the equation  $function(x) = value$  we want to solve on  $]infPoint, supPoint[$

*infPoint* : a real value, the lower bound of the interval where we want to solve the equation

*supPoint* : a real value, the upper bound of the interval where we want to solve the equation

*infValue*: a real value, the value of *function* on the point *infPoint* :  $function(infPoint)$ , must be of opposite sign of *supValue*

*supValue*: a real value, a real value, the value of *function* on the point *supPoint* :  $function(supPoint)$ , must be of opposite sign of *infValue*

**Value** : the result of the root research, in  $]infPoint, supPoint[$

**Details** : Secant might fail and not converge.

## 6.8 ImportanceSampling

This class inherits from Simulation.

**Usage :** *ImportanceSampling(event, importanceDistribution)*

**Arguments :**

*event* : a Event, the event we want to evaluate the probability

*importanceDistribution* : a Distribution, the importance distribution of the Importance Sampling simulation method.

**Some methods :**

*getImportanceDistribution*

**Usage :** *getImportanceDistribution()*

**Arguments :** none

**Value :** a Distribution, the importance distribution of the Importance Sampling simulation method

## 6.9 SimulationResult

**Usage :** structure created by the method `run()` of a `Simulation`, and obtained thanks to the method `getResult()`

**Some methods :**

*getBlockSize*

**Usage :** `getBlockSize()`

**Arguments :** none

**Value :** an integer, the number of terms in the probability simulation estimator grouped together

*getCoefficientOfVariation*

**Usage :** `getCoefficientOfVariation()`

**Arguments :** none

**Value :** a real value, the coefficient of variation of the simulated sample

*getConfidenceLength*

**Usage :** `getConfidenceLength()`

**Arguments :** none

**Value :** a positive real value, the length of any confidence interval equal to the double of the variance of the Monte Carlo estimator

*getOuterSampling*

**Usage :** `getOuterSampling()`

**Arguments :** none

**Value :** an integer, the number of groups of terms in the probability simulation estimator

**Details :** for Monte Carlo, LHS and Importance Sampling methods, the number of evaluations of the limit state function defining the event is : `OuterSampling * BlockSize`

*getProbabilityEstimate*

**Usage :** `getProbabilityEstimate()`

**Arguments :** none

**Value :** a positive real value, the Monte Carlo estimate of the event probability

*getVarianceEstimate*

**Usage :** `getVarianceEstimate()`

**Arguments :** none

**Value :** a positive real value, the variance of the Monte Carlo estimator, equal to the  $Mn(1-Mn) / n$  if  $Mn$  is the Monte Carlo probability estimator and  $n$  the size of the simulated sample

## 7 QuadraticCumul

**Usage :** *QuadraticCumul(randVect)*

**Arguments :** *randVect* : a RandomVector, constraint : this RandomVector must be of type Composite, which means it must have been defined with the second usage of declaration of a RandomVector (from a NumericalMathFunction and an antecedent Distribution)

**Value :** a QuadraticCumul

**Some methods :**

*drawImportanceFactors*

**Usage :** *drawImportanceFactors()*

**Arguments :** none

**Value :** a Graph, the structure containing the pie corresponding to the importance factors of the probabilistic variables

*getCovariance*

**Usage :** *getCovariance()*

**Arguments :** none

**Value :** a CovarianceMatrix, approximation of first order of the covariance matrix of the random vector

*getImportanceFactors*

**Usage :** *getImportanceFactors()*

**Arguments :** none

**Value :** a NumericalPoint, the importance factors of the inputs : only when *randVect* is of dimension 1

*getMeanFirstOrder*

**Usage :** *getMeanFirstOrder()*

**Arguments :** none

**Value :** a NumericalPoint, approximation at the first order of the mean of the random vector

*getMeanSecondOrder*

**Usage :** *getMeanSecondOrder()*

**Arguments :** none

**Value :** a NumericalPoint, approximation at the second order of the mean of the random vector (it requires that the hessian of the NumericalMathFunction has been defined)

## 8 Response Surface Approximation

### 8.1 LinearTaylor

**Usage :** *LinearTaylor*(*center*, *function*)

**Arguments :**

*center* : a NumericalPoint, the point where the Taylor expansion of the function *function* is performed

*function* : a NumericalMathFunction, the function to be approximated.

**Value :** a LinearTaylor

**Some methods :**

*getInputFunction*

**Usage :** *getInputFunction*

**Arguments :** none

**Value :** a NumericalMathFunction, the function *function*

*getName*

**Usage :** *getName*()

**Arguments :** none

**Value :** a string, the name of the LinearTaylor

*getCenter*

**Usage :** *getCenter*()

**Arguments :** none

**Value :** a NumericalPoint, around which the approximation has been made : *center*

*run*

**Usage :** *run*()

**Arguments :** none

**Value :** it performs the linear Taylor expansion around *center* (while this method has not been executed, only *getInputFunction*, *getName* and *setName* methods can be used)

*getConstant*

**Usage :** *getConstant*()

**Arguments :** none

**Value :** a NumericalPoint, the constant vector of the approximation, equal to *function*(*center*)

*getLinear*

**Usage :** *getLinear*()

**Arguments :** none

**Value :** a Matrix, the gradient of the function *function* at the point *center* (the transposition of the jacobian matrix)

*getResponseSurface*

**Usage :** *getResponseSurface*()

**Arguments :** none

**Value :** a NumericalMathFunction, an approximation of the function *function* by a linear Taylor expansion at the *center*

**Links :** see docref\_SurfRep\_Taylor

The methods *getInputFunction*, *getName*, *getCenter* have their associated *setMethod*.

## 8.2 QuadraticTaylor

**Usage :** *QuadraticTaylor(center, function)*

**Arguments :**

*center* : a NumericalPoint, the point where the quadratic Taylor expansion of the function *function* is performed

*function* : a NumericalMathFunction, the function to be approximated : the gradient and hessian of the NumericalMathFunction must be defined.

**Value :** a QuadraticTaylor

**Some methods :**

*getInputFunction*

**Usage :** *getInputFunction*

**Arguments :** none

**Value :** a NumericalMathFunction ,the function *function*

*getName*

**Usage :** *getName()*

**Arguments :** none

**Value :** a string, the name of the Quadratic

*getCenter*

**Usage :** *getCenter()*

**Arguments :** none

**Value :** a NumericalPoint, around which the approximation has been made : *center*

*run*

**Usage :** *run()*

**Arguments :** none

**Value :** it performs the Quadratic Taylor expansion around *center* (while this method has not been executed, only *getInputFunction*, *getName* and *setName* methods can be used)

*getConstant*

**Usage :** *getConstant()*

**Arguments :** none

**Value :** a NumericalPoint, the constant vector of the approximation, equal to *function(center)*

*getLinear*

**Usage :** *getLinear()*

**Arguments :** none



**Value :** a Matrix, the gradient of the function *function* at the point *center* (the transposition of the jacobian matrix)

*getQuadratic*

**Usage :** *getQuadratic()*

**Arguments :** none

**Value :** a SymmetricTensor which contains the 0.5 \* transposition of the hessian values of *function* at *center*

*getResponseSurface*

**Usage :** *getResponseSurface()*

**Arguments :** none

**Value :** a NumericalMathFunction, an approximation of the function *function* by a Quadratic Taylor expansion at *center*

**Links :** see docref\_SurfRep\_Taylor

The methods *getInputFunction*, *getName*, *getCenter* have their associated *setMethod*.

### 8.3 LinearLeastSquares

**Usage :**

*LinearLeastSquares(dataIn, function)*

*LinearLeastSquares(dataIn, dataOut)*

**Arguments :**

*dataIn* : a NumericalSample, the input variables

*function* : a NumericalMathFunction, the function to be approximated

*dataOut* : a NumericalSample, the output variables

**Value :** a LinearLeastSquares, the linear least squares approximation between :

the two samples *dataIn* and *dataOut* in the case of the second usage

the two samples *dataIn* and *function(dataIn)* in the case of the first usage

**Some methods :**

*getInputFunction*

**Usage :** *getInputFunction()*

**Arguments :** none

**Value :** a NumericalMathfunction the *function* parameter in the case of the first usage

*getDataIn*

**Usage :** *getDataIn()*

**Arguments :** none

**Value :** a NumericalSample, the *dataIn* parameter

*getName*

**Usage :** *getName()*

**Arguments :** none

**Value :** a string, the name of the LinearLeastSquares

*run*

**Usage :** *run()*

**Arguments :** none

**Value :** it performs the linear least squares approximation (while this method has not been executed, only *getInputfunctiontion*, *getDataIn*, *getName* and *setName* methods can be used)

*getDataOut*

**Usage :** *getDataIn()*

**Arguments :** none

**Value :** a NumericalSample, it returns the ouput variable :

in the case of the first usage, it corresponds to the values of the function *function* at the input variables *dataIn* : *function(dataIn)*

in the case of the second usage, it corresponds to *dataOut*

#### *getLinear*

**Usage :** *getLinear()*

**Arguments :** none

**Value :** a Matrix, the gradient of the function *function* at the point *center* (the transposition of the jacobian matrix)

#### *getResponseSurface*

**Usage :** *getResponseSurface()*

**Arguments :** none

**Value :** a NumericalMathFunction, an approximation of the function *function* by Linear Least Squares

**Links :** see docref\_\_SurfRep\_\_LeastSquare

The methods *getInputFunction*, *getName*, *getDataIn* have their associated *setMethod*.

## 8.4 QuadraticLeastSquares

**Usage :** *QuadraticLeastSquares(dataIn, function)*  
*QuadraticLeastSquares(dataIn, dataOut)*

**Arguments :**

*dataIn* : a NumericalSample, the input variables  
*function* : a NumericalMathFunction, the function to be approximated  
*dataOut* : a NumericalSample, the output variables

**Value :** a QuadraticLeastSquares, the quadratic least squares approximation between :

the two samples *dataIn* and *dataOut* in the case of the second usage  
the two samples *dataIn* and *function(dataIn)* in the case of the first usage

**Some methods :**

*getInputFunction*

**Usage :** *getInputFunction*  
**Arguments :** none  
**Value :** a NumericalMathFunction, the function *function*

*getName*

**Usage :** *getName()*  
**Arguments :** none  
**Value :** a string, the name of the QuadraticLeastSquares

*getCenter*

**Usage :** *getCenter()*  
**Arguments :** none  
**Value :** a NumericalPoint, around which the approximation has been made : *center*

*run*

**Usage :** *run()*  
**Arguments :** none  
**Value :** it performs the quadratic least squares approximation (while this method has not been executed, only *getInputFunction*, *getName* and *setName* methods can be used)

*getDataOut*

**Usage :** *getDataIn()*  
**Arguments :** none  
**Value :** a NumericalSample, it returns the output variable :  
in the case of the first usage, it corresponds to the values of the function *function* at the input variables *dataIn* : *function(dataIn)*

in the case of the second usage, it corresponds to *dataOut*

*getConstant*

**Usage :** *getConstant()*

**Arguments :** none

**Value :** a NumericalPoint, the constant vector of the approximation, equal to *function(center)*

*getLinear*

**Usage :** *getLinear()*

**Arguments :** none

**Value :** a Matrix, the linear matrix of the approximation

*getQuadratic*

**Usage :** *getQuadratic()*

**Arguments :** none

**Value :** a SymmetricTensor, the quadratic term of the approximation

*getResponseSurface*

**Usage :** *getResponseSurface()*

**Arguments :** none

**Value :** a NumericalMathFunction, an approximation of the function *function* by Quadratic Least Squares

**Links :** see `docref_SurfRep_LeastSquare`

The methods *getInputFunction*, *getName*, *getDataIn* have their associated *setMethod*.

## 9 Graphs

### 9.1 Graph

The class Graph is the structure which contains :

- the drawable elements (may be several drawables elements) : class Drawable
- the graphical context : the potential axes and labels, the bounding box, the global title, the global legend and its position

**Usage :**

*Graph(title, xTitle, yTitle, showAxes)*

*Graph(title, xTitle, yTitle, showAxes, legendPosition)*

*Graph(title, xTitle, yTitle, showAxes, legendPosition, legendFontSize)*

**Arguments :**

*title* : a String, the title of the graph

*xTitle* : a String, the legend of the X axe

*yTitle* : a String, the legend of the Y axe

*showAxes* : a boolean which indicates if the axes are drawn (yes = 1, no = 0)

*legendPosition* : a String which indicates the position of the legend. If *legendPosition* is not specified, the Graph has no legend

*legendFontSize* : an interger, the font size of the legend. If not specified, the default width will be used

**Some methods :**

*addDrawable*

**Usage :** *addDrawable(aDrawable)*

**Arguments :** *aDrawable* : a Drawable, a drawable element we want to add on the graph

**Value :** none, it adds the new graph on the first one, with its legend. It keeps the graphical context of the first graph

*draw*

**Usage :**

*draw(path, file, width, height)*

*draw(file, width, height)*

*draw(file)*

**Arguments :**

*path* : a String which indicates the adress where the created file will be put. When not specified, the files is created in the cururent repertory

*file* : a String wich indicates the name of the created file (without the suffixe). The files created will be file.png and file.ps

*width, height* : two real positive values, number of pixels fixing the width and the height of the graph. When not specified, the couple (640,480) is taken into account

**Value :** none : it generates the files file.png and file.ps

*getAxes*

**Usage :** *getAxes()*

**Arguments :** none

**Value :** a boolean which indicates if the axes are drawn (yes = 1, no = 0)

*getBitmap*

**Usage :** *getBitmap()*

**Arguments :** none

**Value :** a String, the adress of the file file.png created by the method draw

*getBoundingBox*

**Usage :** *getBoundingBox()*

**Arguments :** none

**Value :** a NumericalPoint of dimension 4, the bounding box of the drawable element, wich is a rectangle determined by its low and left corner (P1) and its high and right corner (P2). The BoundingBox is (XP1,YP1, XP2, YP2).

*getDrawables*

**Usage :** *getDrawables()*

**Arguments :** none

**Value :** a DrawableCollection, the collection of the Drawables included in the graph

*getFileName*

**Usage :** *getFileName()*

**Arguments :** none

**Value :** a String, the name of the files containing the graph

*getLegendFontSize*

**Usage :** *getLegendFontSize()*

**Arguments :** none

**Value :** a positive real, the legend font size

*getLegendPosition*

**Usage :** *getLegendPosition()*

**Arguments :** none

**Value :** a String, the position of the legend.

*getPath*

**Usage :** *getPath()*

**Arguments :** none

**Value :** a String, the adress where the files file.png and file.ps are put

*getPostscript*

**Usage :** *getPostscript()*

**Arguments :** none

**Value :** a String, the adress of the file file.ps created by the method .draw()

*getTitle*

**Usage :** *getTitle()*

**Arguments :** none

**Value :** a String, the title of the graph

*getXTitle*

**Usage :** *getXTitle()*

**Arguments :** none

**Value :** a String, the title of the X axe

*getYTitle*

**Usage :** *getYTitle()*

**Arguments :** none

**Value :** a String, the title of the Y axe

*setBoundingBox*

**Usage :** *setBoundingBox(myBoundingBox)*

**Arguments :** *myBoundingBox* : a BoundingBox, which is a Numericalpoint(4) composed by  $[x_{min}, x_{max}, y_{min}, y_{max}]$  if we want to impose the x-range to  $[x_{min}, x_{max}]$  and the y-range to  $[y_{min}, y_{max}]$ .

**Value :** none

The methods *getAxes*, *getDrawables*, *getLegendPosition*, *getTitle*, *getXTitle*, *getYTitle* have their corresponding *setMethod*.

Here is the list of legend positions accepted by Open TURNS : "bottomright", "bottom", "bottomleft", "left", "topleft", "topright", "right", "center".



## 9.2 Drawable

A Drawable is a drawable element described by :

- its data,
- their attributes : color, line stype, point style, fill style
- the specific legend of the drawable element.

**Usage :** *Drawable(drawableImplementation)*

**Arguments :** *drawableImplementation* : a DrawableImplementation, the implementation of Drawable, which is *Curve*, *Cloud*, *BarPlot*, *Staircase*, *Pie*

**Some methods :**

*getBoundingBox*

**Usage :** *getBoundingBox()*

**Arguments :** none

**Value :** a NumericalPoint of dimension 4, the bounding bow of the drawable element, wich is a rectangle determined by its low and left corner (P1)and its high and right corner (P2). The BoundingBox is (XP1,YP1, XP2, YP2).

*getColor*

**Usage :** *getColor()*

**Arguments :** none

**Value :** a String which describes the color of the lines within the drawable element none

*getData*

**Usage :** *getData()*

**Arguments :** none

**Value :** a NumericalSample, from which the Drawable is built

*getFillStyle*

**Usage :** *getFillStyle()*

**Arguments :** none

**Value :** a String which describes the fill style of the surfaces within the drawable element

*getLabels*

**Usage :** *getLabels()*

**Arguments :** none

**Value :** a Description, the labels of both axes

*getLegendName*

**Usage :** *getLegendName()*

**Arguments :** none

**Value :** a String which is the legend of the drawable element

*getLineStyle*

**Usage :** *getLineStyle()*

**Arguments :** none

**Value :** a String which describes the style of the lines within the drawable element

*getLineWidth*

**Usage :** *getLineWidth()*

**Arguments :** none

**Value :** an interger, the width of the line included in the Drawable (if such the case)

*getPointCode*

**Usage :** *getPointCode()*

**Arguments :** none

**Value :** an integer which describes the style of the points within the drawable element

*getPointStyle*

**Usage :** *getPointStyle()*

**Arguments :** none

**Value :** a string which describes the style of the points within the drawable element

All the methods *getColor*, *getFillStyle*, *getLineStyle*, *getPointCode* and *getPointStyle* have their corresponding *setMethod*.

Here is the list of codes, styles, width accepted by Open TURNS :

- map matching keys with R codes for point symbols :

Point Style	Point Code
square	0
circle	1
triangleup	2
plus	3
times	4
diamond	5
triangledown	6
star	8
fsquare	15
fcircle	16
ftriangleup	17
fdiamond	18
bullet	20

- possible colors : "green", "red", "blue", "yellow", "darkblue", "orange", "lightgreen", "darkcyan", "cyan", "magenta", "darkgreen", "violet", "brown", "darkred", "pink", "ivory", "gold", "darkgrey", "grey", "white", "aliceblue", "antiquewhite", "antiquewhite1", "antiquewhite2", "antiquewhite3", "antiquewhite4", "aquamarine", "aquamarine1", "aquamarine2", "aquamarine3", "aquamarine4", "azure", "azure1", "azure2", "azure3", "azure4", "beige", "bisque", "bisque1", "bisque2", "bisque3", "bisque4", "black", "blanchedalmond", "blue1", "blue2", "blue3", "blue4", "blueviolet", "brown1", "brown2", "brown3", "brown4", "burlywood", "burlywood1", "burlywood2", "burlywood3", "burlywood4", "cadetblue", "cadetblue1", "cadetblue2", "cadetblue3", "cadetblue4", "chartreuse", "chartreuse1", "chartreuse2", "chartreuse3", "chartreuse4", "chocolate", "chocolate1", "chocolate2", "chocolate3", "chocolate4", "coral", "coral1", "coral2", "coral3", "coral4", "cornflowerblue", "cornsilk", "cornsilk1", "cornsilk2", "cornsilk3", "cornsilk4", "cyan1", "cyan2", "cyan3", "cyan4", "darkgoldenrod", "darkgoldenrod1", "darkgoldenrod2", "darkgoldenrod3", "darkgoldenrod4", "darkgray", "darkkhaki", "darkmagenta", "darkolivegreen", "darkolivegreen1", "darkolivegreen2", "darkolivegreen3", "darkolivegreen4", "darkorange", "darkorange1", "darkorange2", "darkorange3", "darkorange4", "darkorchid", "darkorchid1", "darkorchid2", "darkorchid3", "darkorchid4", "darksalmon", "darkseagreen", "darkseagreen1", "darkseagreen2", "darkseagreen3", "darkseagreen4", "darkslateblue", "darkslategray", "darkslategray1", "darkslategray2", "darkslategray3", "darkslategray4", "darkslategrey", "darkturquoise", "darkviolet", "deeppink", "deeppink1", "deeppink2", "deeppink3", "deeppink4", "deepskyblue", "deepskyblue1", "deepskyblue2", "deepskyblue3", "deepskyblue4", "dimgray", "dimgrey", "dodgerblue", "dodgerblue1", "dodgerblue2", "dodgerblue3", "dodgerblue4", "firebrick", "firebrick1", "firebrick2", "firebrick3", "firebrick4", "floralwhite", "forestgreen", "gainsboro", "ghostwhite", "gold1", "gold2", "gold3", "gold4", "goldenrod", "goldenrod1", "goldenrod2", "goldenrod3", "goldenrod4", "gray", "gray0", "gray1", "gray2", "gray3", "gray4", "gray5", "gray6", "gray7", "gray8", "gray9", "gray10", "gray11", "gray12", "gray13", "gray14", "gray15", "gray16", "gray17", "gray18", "gray19", "gray20", "gray21", "gray22", "gray23", "gray24", "gray25", "gray26", "gray27", "gray28", "gray29", "gray30", "gray31", "gray32", "gray33", "gray34", "gray35", "gray36", "gray37", "gray38", "gray39", "gray40", "gray41", "gray42", "gray43", "gray44", "gray45", "gray46", "gray47", "gray48", "gray49", "gray50", "gray51", "gray52", "gray53", "gray54", "gray55", "gray56", "gray57", "gray58", "gray59", "gray60", "gray61", "gray62", "gray63", "gray64", "gray65", "gray66", "gray67", "gray68", "gray69", "gray70", "gray71", "gray72", "gray73", "gray74", "gray75", "gray76", "gray77", "gray78", "gray79", "gray80", "gray81", "gray82", "gray83", "gray84", "gray85", "gray86", "gray87", "gray88", "gray89", "gray90", "gray91", "gray92", "gray93", "gray94", "gray95", "gray96", "gray97", "gray98", "gray99", "gray100", "green1", "green2", "green3", "green4", "greenyellow", "grey0", "grey1", "grey2", "grey3", "grey4", "grey5", "grey6", "grey7", "grey8", "grey9", "grey10", "grey11", "grey12", "grey13", "grey14", "grey15", "grey16", "grey17", "grey18", "grey19", "grey20", "grey21", "grey22", "grey23", "grey24", "grey25", "grey26", "grey27", "grey28", "grey29", "grey30", "grey31", "grey32", "grey33", "grey34", "grey35", "grey36", "grey37", "grey38", "grey39", "grey40", "grey41", "grey42", "grey43", "grey44", "grey45", "grey46", "grey47", "grey48", "grey49", "grey50", "grey51", "grey52", "grey53", "grey54", "grey55", "grey56", "grey57", "grey58", "grey59", "grey60", "grey61", "grey62", "grey63", "grey64", "grey65", "grey66", "grey67", "grey68", "grey69", "grey70", "grey71", "grey72", "grey73", "grey74", "grey75", "grey76", "grey77", "grey78", "grey79", "grey80", "grey81", "grey82", "grey83", "grey84", "grey85", "grey86", "grey87", "grey88", "grey89", "grey90", "grey91", "grey92", "grey93", "grey94", "grey95", "grey96", "grey97", "grey98", "grey99", "grey100", "honeydew", "honeydew1", "honeydew2", "honeydew3", "honeydew4", "hotpink", "hotpink1", "hotpink2", "hotpink3", "hotpink4", "indianred", "indianred1", "indianred2", "indianred3", "indianred4", "ivory1", "ivory2", "ivory3", "ivory4", "khaki", "khaki1", "khaki2", "khaki3", "khaki4", "lavender", "lavenderblush", "lavenderblush1", "lavenderblush2", "lavenderblush3", "lavenderblush4", "lawngreen", "lemonchiffon", "lemonchiffon1", "lemonchiffon2", "lemonchiffon3", "lemonchiffon4", "lightblue", "lightblue1", "lightblue2", "lightblue3", "lightblue4", "lightcoral", "lightcyan", "lightcyan1", "lightcyan2", "lightcyan3", "lightcyan4",

"lightgoldenrod", "lightgoldenrod1", "lightgoldenrod2", "lightgoldenrod3", "lightgoldenrod4", "lightgoldenrodyellow", "lightgray", "lightgrey", "lightpink", "lightpink1", "lightpink2", "lightpink3", "lightpink4", "lightsalmon", "lightsalmon1", "lightsalmon2", "lightsalmon3", "lightsalmon4", "lightseagreen", "lightskyblue", "lightskyblue1", "lightskyblue2", "lightskyblue3", "lightskyblue4", "lightslateblue", "lightslategray", "lightslategrey", "lightsteelblue", "lightsteelblue1", "lightsteelblue2", "lightsteelblue3", "lightsteelblue4", "lightyellow", "lightyellow1", "lightyellow2", "lightyellow3", "lightyellow4", "limegreen", "linen", "magenta1", "magenta2", "magenta3", "magenta4", "maroon", "maroon1", "maroon2", "maroon3", "maroon4", "mediumaquamarine", "mediumblue", "mediumorchid", "mediumorchid1", "mediumorchid2", "mediumorchid3", "mediumorchid4", "mediumpurple", "mediumpurple1", "mediumpurple2", "mediumpurple3", "mediumpurple4", "mediumseagreen", "mediumslateblue", "mediumspringgreen", "mediumturquoise", "mediumvioletred", "midnightblue", "mintcream", "mistyrose", "mistyrose1", "mistyrose2", "mistyrose3", "mistyrose4", "moccasin", "navajowhite", "navajowhite1", "navajowhite2", "navajowhite3", "navajowhite4", "navy", "navyblue", "oldlace", "olivedrab", "olivedrab1", "olivedrab2", "olivedrab3", "olivedrab4", "orange1", "orange2", "orange3", "orange4", "orangered", "orangered1", "orangered2", "orangered3", "orangered4", "orchid", "orchid1", "orchid2", "orchid3", "orchid4", "palegoldenrod", "palegreen", "palegreen1", "palegreen2", "palegreen3", "palegreen4", "paleturquoise", "paleturquoise1", "paleturquoise2", "paleturquoise3", "paleturquoise4", "palevioletred", "palevioletred1", "palevioletred2", "palevioletred3", "palevioletred4", "papayawhip", "peachpuff", "peachpuff1", "peachpuff2", "peachpuff3", "peachpuff4", "peru", "pink1", "pink2", "pink3", "pink4", "plum", "plum1", "plum2", "plum3", "plum4", "powderblue", "purple", "purple1", "purple2", "purple3", "purple4", "red1", "red2", "red3", "red4", "rosybrown", "rosybrown1", "rosybrown2", "rosybrown3", "rosybrown4", "royalblue", "royalblue1", "royalblue2", "royalblue3", "royalblue4", "saddlebrown", "salmon", "salmon1", "salmon2", "salmon3", "salmon4", "sandybrown", "seagreen", "seagreen1", "seagreen2", "seagreen3", "seagreen4", "seashell", "seashell1", "seashell2", "seashell3", "seashell4", "sienna", "sienna1", "sienna2", "sienna3", "sienna4", "skyblue", "skyblue1", "skyblue2", "skyblue3", "skyblue4", "slateblue", "slateblue1", "slateblue2", "slateblue3", "slateblue4", "slategray", "slategray1", "slategray2", "slategray3", "slategray4", "slategrey", "snow", "snow1", "snow2", "snow3", "snow4", "springgreen", "springgreen1", "springgreen2", "springgreen3", "springgreen4", "steelblue", "steelblue1", "steelblue2", "steelblue3", "steelblue4", "tan", "tan1", "tan2", "tan3", "tan4", "thistle", "thistle1", "thistle2", "thistle3", "thistle4", "tomato", "tomato1", "tomato2", "tomato3", "tomato4", "turquoise", "turquoise1", "turquoise2", "turquoise3", "turquoise4", "violetred", "violetred1", "violetred2", "violetred3", "violetred4", "wheat", "wheat1", "wheat2", "wheat3", "wheat4", "whitesmoke", "yellow1", "yellow2", "yellow3", "yellow4", "yellowgreen"

- line styles : "blank", "solid", "dashed", "dotted", "dotdash", "longdash", "twodash"
- some fill styles (patterns) : "solid", "shaded"

The default values are the following ones :

- *Color* = "blue"
- *SurfaceColor* = "white"
- *FillStyle* = "solid"
- *PointStyle* = "plus"
- *LineWidth* = 1
- *LineStyle* = "solid"
- *Pattern* = "s"

### 9.3 Curve

It inherits from the methods of the Drawable class.

#### Usage :

*Curve(data, legend)*

*Curve(data, color, lineStyle, lineWidth, legend)*

#### Arguments :

*data* : a NumericalSample, the points from which the curve is built, must be of dimension 2

*legend* : a String, the legend

*color* : a String, the color of the curve

*lineStyle* : a String, the style of the curve

*lineWidth* : an integer, the line width of the curve

#### Some methods :

*isConformData*

**Usage :** *isConformData(data)*

**Arguments :** *data* : a NumericalSample

**Value :** a boolean which indicates if the type of data is conform to the type of the drawable (here a Curve) : a NumericalSample of dimension 2

*getData*

**Usage :** *getData()*

**Arguments :** none

**Value :** a NumericalSample of dimension 2, the data from which the curve is built

*getLineWidth*

**Usage :** *getLineWidth()*

**Arguments :** none

**Value :** an integer, the line width of the curve

All the methods *getColor*, *getLegendName*, *getLineStyle* and *getLineWidth* have their corresponding *setMethod*.

## 9.4 Cloud

It inherits from the methods of the Drawable class.

### Usage :

*Cloud(data, legend)*

*Cloud(data, color, pointStyle, legend)*

### Arguments :

*data* : a NumericalSample, the points from which the cloud is built, must be of dimension 2

*legend* : a String, the legend

*color* : a String, the color of the curve . If not specified, by default equal to "blue"

*pointStyle* : a String, the style of the points. If not specified, by default equal to "plus"

### Some methods :

*isConformData*

**Usage :** *isConformData(data)*

**Arguments :** *data* : a NumericalSample

**Value :** a boolean which indicates if the type of data is conform to the type of the drawable : a NumericalSample of dimension 2

*getData*

**Usage :** *getData()*

**Arguments :** none

**Value :** a NumericalSample of dimension 2, the data from which the cloud is built

All the methods *getColor*, *getLegendName*, *getPointCode* and *getPointStyle* have their corresponding *setMethod*.

## 9.5 BarPlot

It inherits from the methods of the Drawable class.

**Usage :**

*BarPlot(data, origin, legend)*

*BarPlot(data, origin, color, fillStyle, lineStyle, legend)*

**Arguments :**

*data* : a NumericalSample, the data from which the BarPlot is built, must be of dimension 2 : the discontinuous points and their corresponding height

*origin* : a real value which is where the BarPlot begins

*legend* : a String, the legend

*color* : a String, the color of the curve . If not specified, by default equal to "blue"

*lineStyle* : a String, the style of the curve. If not specified, by default equal to "solid"

*fillStyle* : a String, the fill style of the surfaces. If not specified, by default equal to "solid"

**Some methods :**

*getData*

**Usage :** *getData()*

**Arguments :** none

**Value :** a NumericalSample, of dimension 2, giving the discontinuous points and their corresponding height

*getOrigin*

**Usage :** *getOrigin()*

**Arguments :** none

**Value :** a real value which is where the BarPlot begins

*isConformData*

**Usage :** *isConformData(data)*

**Arguments :** *data* : a NumericalSample

**Value :** a boolean which indicates if the type of data is conform to the type of the drawable (here a BarPlot) : a NumericalSample, of dimension 2

All the methods *getColor*, *getLegendName*, *getOrigin*, *getFillStyle* and *getLineStyle* have their corresponding *setMethod*.

## 9.6 Staircase

It inherits from the methods of the Drawable class.

### Usage :

*StairCase(data, legend)*

*StairCase(data, color, lineStyle, pattern, legend)*

### Arguments :

*data* : a NumericalSample, the points from which the Staircase is built, must be of dimension 2 : the discontinuous points and their corresponding height

*legend* : a String, the legend

*color* : a String, the color of the curve If not specified, by default equal to "blue"

*lineStyle* : a String, the style of the curve. If not specified, by default equal to "solid"

*pattern* : a String, the pattern of the surfaces. If not specified, by default equal to "s"

### Some methods :

*getData*

**Usage :** *getData()*

**Arguments :** none

**Value :** a NumericalSample, of dimension 2, giving the discontinuous points and their corresponding height

*isConformData*

**Usage :** *isConformData(data)*

**Arguments :** *data* : a NumericalSample

**Value :** a boolean which indicates if the type of data is conform to the type of the drawable (here a Staircase) : a NumericalSampl of dimension 2, giving the discontinuous points and their corresponding height

*getPattern*

**Usage :** *getPattern()*

**Arguments :** none

**Value :**

All the methods *getColor*, *getLegendName*, *getLineStyle* and *getPattern* have their corresponding *setMethod*.



## 9.7 Pie

It inherits from the methods of the Drawable class.

### Usage :

*Pie(data)*

*Pie(data, labels, center, radius, palette)*

### Arguments :

*data* : a NumericalSample, of dimension 1, giving the percentiles of the pie

*labels* : a StringCollection, the names of each group. If not specified, by default equal to the description of the probabilistic input vector

*center* : a NumericalPoint, the center of the pie inside the bounding box. If not specified, by default equal to (0,0)

*radius* : a real positive value, the radius of the pie. If not specified, by default equal to 1

*palette* : a StringCollection, the names of the colors. If not specified, colors are successively taken from the list given below, in the same order

### Some methods :

*getCenter*

**Usage :** *getCenter()*

**Arguments :** none

**Value :** a NumericalPoint, the center of the pie inside the bounding box

*getData*

**Usage :** *getData()*

**Arguments :** none

**Value :** a NumericalSample of dimension 1, giving the percentiles of the pie

*getLabels*

**Usage :** *getLabels()*

**Arguments :** none

**Value :** a StringCollection, the names of each group

*getPalette*

**Usage :** *getPalette()*

**Arguments :** none

**Value :** a StringCollection, the names of the colors used for the pie

*getRadius*

**Usage :** *getRadius()*

**Arguments :** none

**Value :** a real positive value, the radius of the pie

*isConformData*

**Usage :** *isConformData(data)*

**Arguments :** *data* : a NumericalSample

**Value :** a boolean which indicates if the type of data is conform to the type of the drawable (here a Pie) : a NumericalSample of dimension 1

*getPattern*

**Usage :** *getPattern()*

**Arguments :** none

**Value :**

All the methods *getColor*, *getLegendName*, *getLineStyle* and *getPattern* have their corresponding *setMethod*.

## 9.8 Show

The command *Show* enables to visualise a Graph within the TUI without creating the files .EPS, .PNG or .FIG.

**Usage :** *Show(graph)*

**Arguments :** *graph* : a Graph

**Value :** It shows the graph within the TUI without saving it in any file.

**Details :** For example, *Show(myDistribution.drawPDF())* where *myDistribution* is a Distribution.