

MANUAL

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RAVEN User Manual

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RAVEN User Manual

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RAVEN is a bird

1 Introduction

TO BE ADDED

2 Raven Input Structure

The RAVEN code does not have a fixed calculation flow, since all its basic objects can be combined in order to create a user-defined calculation flow. Thus, its input (xml) is organized in different xml blocks, each with a different functionality. The main input blocks are as follows:

- **Simulation:** The simulation block is the one that has inside the entire input, all the following blocks fit inside the simulation block;
- **RunInfo:** Block in which the calculation settings are specified (number of parallel simulations, etc.);
- **Distributions:** Distributions' container;
- **Samplers:** Exploration of the uncertain domain strategy specification;
- **Functions:** External functions container;
- **Models:** Models' specifications (e.g. Codes,ROM,etc.);
- **Steps:** Place where the single basic objects get combined;
- **Datas:** Internal Data object block;
- **Databases:** Databases block;
- **OutStream system:** Visualization and Printing system block.

Each of these blocks are explained in dedicated sections in the following chapters.

3 RunInfo

The *RunInfo* block is the place where the user specifies how the calculation needs to be performed. In this input block, several settings can be inputted, in order to define how to drive the calculation and set up, when needed, particular settings for the machine the code needs to run on (queue system if not PBS, etc.). In the following, all the keywords are explained in detail.

- *< WorkingDir >*, in this block the user needs to specify the absolute or relative (with respect to the location where RAVEN is run from) path to a directory that is going to be used to store all the results of the calculations and where RAVEN looks for the files specified in the block *< Files >*;
- *< ParallelCommand >*: the command that should be used to submit jobs in parallel
- *< queuingSoftware >*: queuing software name
- *< ThreadingCommand >*: The command should be used to submit multi-threaded jobs????
- *< NumThreads >*: number of threads by run
- *< NumNode >*: number of nodes used in the HPC
- *< procByNode > **: number of processors used in a single node
- *NumCoresUsed > **: total number of cores used
- *< NumMPI > **: number of MPI processes by run
- *< MaxLogFileSize > * :******
- *< precommand >*: add before of the command that is used to run the external model
- *< postcommand >*: add after of the command that is used to run the external model
- *< deleteOutExtension >*: if a run of an external model has not failed delete the output files with the listed extension

- *< delSucLogFiles >*: if a run of an external model has not failed delete the log files with the listed extension
- *< Files >*: these are the paths to the files required by the code, string from the working directory
- *< Sequence >*: ordered list of the step name that the simulation will run
- *< mode >*: these refers to the way in which the parallel environment is set up (currently the only modes supported are pbsdsh and mpi). When the user is running the code on his own machine *no mode should be set*. On the contrary, when the user wants to run the code on the cluster, he should specify the mode. Additional info can be found in the readme file.
- *< batchSize >*: number of contemporaneous processors used when the code is running on the cluster. It is good practice to try to run the input a first time using a batchSize=1 to see if there are no faults inside the RAVEN input and/or any code that is coupled with RAVEN. If there are no faults then the user can input batchSize= to as many processors are needed to be used.
- *< expectedTime >**: how much time the simulation is expected to run. After this period of time the cluster will automatically stop the simulation (even if the simulation is not completed) so be sure that the time is overestimated.
- *< DefaultInputFile >*: default input file to be read*****
- *< CustomMode >*: *****

The example:

```
<RunInfo>
  <WorkingDir>externalModel</WorkingDir>
  <Files>lorentzAttractor.py</Files>
  <Sequence>MonteCarlo</Sequence>
  <batchSize>100</batchSize>
  <NumThreads>4</NumThreads>
  <mode>mpi</mode>
  <NumMPI>2</NumMPI>
</RunInfo>
```

Specifies the working directory (WorkingDir) where are located the files necessary (Files) to run a series of 100 (batchSize) Monte-Carlo calculations (Sequence). MPI (mode) mode is used along with 4 threads (NumThreads) and 2 mpi process per run (NumMPI).

4 Distributions

All of the probability distributions functions could be truncated by using:

```
<low> * * * </low>  
<hi> * * * </hi>
```

In case the probability distribution function demanded specific boundary values, those are set by default, and the user can change them.

- The **Bernoulli** distribution is a discrete distribution of the outcome of a single experiment with success probability p .

```
<Bernoulli name='...'>  
<p> * * * </p>
```

- The **Beta** distribution is a continuous distribution. It has four parameters, low for the lowest value, $high$ for the highest value, $alpha$ and $beta$ for the shape parameters.

```
<Beta name='...'>  
<low> * * * </low>  
<hi> * * * </hi>  
<alpha> * * * </alpha>  
<beta> * * * </beta>
```

- The **Binomial** distribution is a discrete distribution of the outcome of n experiments with each experiment having the success probability p .

```
<Binomial name='...'>  
<n> * * * </n>  
<p> * * * </p>
```

- The **Exponential** distribution is a continuous distribution of which can be used to model the time between independent events that happen at a constant average time. It uses the rate parameter $lambda$.

```
<Exponential name='...'>  
<lambda> * * * </lambda>
```

- The **Gamma** distribution is a continuous distribution. It has three parameters, *low* for the lowest value, *alpha* is the shape parameter, and *beta* which is 1/scale or the inverse scale parameter.

```
<Gamma name='...'>
<low>***</low>
<alpha>***</alpha>
<beta>***</beta>
```

- The **Logistic** distribution is a continuous distribution similar to the normal distribution with a CDF that is an instance of a logistic function. It has two parameters, *location* which is the most common value and the center, and *scale* which determines the shape.

```
<Logistic name='...'>
<location>***</location>
<scale>***</scale>
```

- The **LogNormal** distribution is a continuous distribution with the logarithm of the random variable being normally distributed. It has two parameters, *mean* which is the expected value, and *sigma* which is the standard deviation.

```
<LogNormal name='...'>
<mean>***</mean>
<sigma>***</sigma>
```

- The **Normal** distribution (or Gaussian) distribution is a continuous distribution which because of the central limit theorem, the mean of many distributions approximates a normal distribution. It has two parameters, *mean* which is the middle value, and *sigma* which is the standard deviation.

```
<Normal name='...'>
<mean>***</mean>
<sigma>***</sigma>
```

- The **Poisson** distribution is a discrete distribution that expresses the probability of the number of events occurring in a fixed period of time. It has one parameter, *mu* the mean rate of events/time.

```
<Poisson name='...'>
<mu>***</mu>
```

- The **Triangular** distribution is a continuous distribution that has a triangular shape for the PDF. The peak falls at the *apex* and the values run from *min* to *max*.

```
<Triangular name='...'>
<apex>***</apex>
<min>***</min>
<max>***</max>
```

- The **Uniform** distribution is a continuous distribution with a rectangular shaped PDF. It goes from *low* to *high*.

```
<Uniform name='...'>
<low>***</low>
<hi>***</hi>
```

- The **Weibull** distribution is a continuous distribution that can be used in failure analysis. It takes two parameters, *k* or the shape parameter, and *lambda* or the scale parameter.

```
<Weibull name='...'>
<lambda>***</lambda>
<k>***</k>
```


5 Samplers

Samplers own the sampling strategy (Type) and they generate the input values using the associate distribution. They do not have distributions inside.

There are different kind of samplers:

- **MonteCarlo**
- **Grid**
- **LHS**
- **Adaptive**
- **DynamicEventTree**
- **AdaptiveDynamicEventTree**

For MonteCarlo method:

- **MonteCarlo**
 - name: name of sampling method used
 - limit: number of samplings the MonteCarlo method will use
 - initial seed: (optional) initial number of the iterations
- variable
 - name: name of the variable that the code will do the sampling of
- distribution:
 - what kind of distribution the variable follows (it is chosen from the distribution block)

Example:

```

<MonteCarlo name='MC_Sampler' limit='1000'>
  <variable name='pressure'>
    <distribution>***</distribution>
  </variable>
</MonteCarlo>

```

In the example the code is sampling 1000 times (limit) the variable "pressure" with a distribution taken from the distribution block.

For the Grid method:

- **Grid**

- name: name of the sampling method used
- initial seed (optional)

- variable

- name: name of the variable that the code will do the sampling of

- distribution: what kind of distribution the variable follows (it is chosen from the distribution block)

- grid

- type

- * value
- * CDF

- construction

- * equal

- the size of the step, given in the input node, is the same for all steps
- requires lowerbound or upperbound
- requires steps

- * custom

- no lowerbound or upperbound
- no number of steps
- in the input node it is necessary to input each step of the grid separated by a space

- lowerbound or upperbound
- steps
- input node

Example:

```
<Grid name='Grid_Sampler'>
  <variable name='pressure'>
    <distribution>***</distribution>
    <grid type='value' construction='equal' steps='100'
      lowerBound='1.0'>0.2</grid>
  </variable>
</Grid>
```

In the example the code is sampling the variable "pressure" with a distribution "***" which was chosen by the distribution block. The grid ranges from 1.0 (lowerbound) to 21.0 (100 steps of equal size of 0.2).

For the LHS method:

- **LHS**

- name: name of the sampling method used
- initial seed (optional)
- variable
 - name: name of the variable that the code will do the sampling of
- distribution: what kind of distribution the variable follows (it is chosen from the distribution block)
- grid
 - type
 - * value
 - * CDF
 - construction
 - * equal

- the size of the step, given in the input node, is the same for all steps
- requires lowerbound or upperbound
- requires steps
- * custom
 - no lowerbound or upperbound
 - no number of steps
 - in the input node it is necessary to input each step of the grid separated by a space
- lowerbound or upperbound
- steps
- input node

Example:

```
<LHS name='***' initial_seed='***'>
  <variable name='***'>
    <distribution>***</distribution>
    <grid type='***' construction='***' steps='***'
      lowerBound='***'>***</grid>
  </variable>
</LHS>
```

The input structure is the same as the **Grid** input structure

For the Adaptive method:

- **Adaptive**
 - name
 - initial seed (optional)
- **Convergence**
 - limit: 'Integer'
 - persistence: 'Integer'
 - weight='probability' or 'None':

- subGridTol='None' or 'Float' :This is the tolerance used to construct the testing sub grid
 - forcelteration='True' or 'False': this flag control if at least a self.limit number of iteration should be done
- variable
 - distribution

```

<Adaptive name='***' initial_seed='***'>
  <Convergence limit='***' persistence='***' weight='***' subGridTol='***'
    forcelteration='***'>***</Convergence>
  <variable name='***'>
    <distribution>***</distribution>
  </variable>
</Adaptive>

```

For DynamicEventTree: ***

For AdaptiveDynamicEventTree: ***

6 Functions

This module contains interfaces to import external functions

- **External**
 - name=name of the function
 - file=file name where the function is
- variable
 - type=numpy.float64
 - variable inside the function that has been defined

```
<Functions>  
<External name='***' file='***'>  
  <variable type='***'>***</variable>  
</External>  
</Functions>
```

7 Models

The xml section "models" contains the information regarding the code employed in the analysis (e.g., RAVEN/RELAP-7, RELAP-5 or an external model). A model is something that given an input will return an output reproducing some physical model it could be as complex as a stand alone code, a reduced order model trained somehow or something externally build and imported by the user. The available models are:

Dummy: it is a dummy model that just return the effect of the sampler. The values reported as input in the output are the output of the sampler and the output is the counter of the performed sampling

ROM: ROM stands for Reduced Order Model. All the models here, first learn than predict the outcome

ExternalModel: this model allows to interface with an external python module

Code: generic class that imports an external code into the framework

Projector: generic data manipulator

PostProcessor: an Action System. All the models here, take an input and perform an action

7.1 Dummy

Description

Summary

Example

7.2 ROM

Description

Summary

- name: name of the ROM model
- subType: 'SciKitLearn'. Imports the libraries from scikitlearn
- Features: input the variables set in the Samplers block separated by a comma
- SKLtype: input a model from <http://scikit-learn.org/stable/modules/classes.html> (LinearRegression in the example)
- Target: Input a set of variables inside the output space which the ROM has to be developed for.
- Parameters: From the SKLtype class are defined all the parameters required (*fit_intercept* and *normalize* in the exemple)

Example

```
<Models>
<ROM name='***' subType='***'>
  <Features>***, ***</Features>
  <SKLtype>linear_model | LinearRegression</SKLtype>
  <Target>***</Target>
  <fit_intercept>***</fit_intercept>
  <normalize>***</normalize>
</ROM>
```

7.3 External Model

Description

Summary

Example As an example we use the external model shown in `lorenzAttractor.py` which, given the 3-dimensional initial coordinates (x0, y0, z0), calculate the trajectory of a Lorentz attractor in the time interval [0.0, 0.03] seconds. We want to perform sampling of the 3-dimensional initial conditions of the attractor using classical Monte-Carlo sampling. The user is required to specify:

- the initialize function: `def initialize(self,runInfoDict,inputFiles)`

- the function which create a new input: `def createNewInput(self,myInput,samplerType,**Kwargs)`
- the function which perform the actual calculation: `def run(self,Input)`

```
def initialize(self,runInfoDict,inputFiles):
    self.SampledVars = None
    self.sigma = 10.0
    self.rho = 28.0
    self.beta = 8.0/3.0
    return

def createNewInput(self,myInput,samplerType,**Kwargs):
    return Kwargs['SampledVars']

def run(self,Input):
    ...
```

<Models>

```
<ExternalModel name='PythonModule' subType='
ModuleToLoad='externalModel/lorentzAttractor'>
  <variable type='float'>sigma</variable>
  <variable type='float'>rho</variable>
  <variable type='float'>beta</variable>
  <variable type='numpy.ndarray'>x</variable>
  <variable type='numpy.ndarray'>y</variable>
  <variable type='numpy.ndarray'>z</variable>
  <variable type='numpy.ndarray'>time</variable>
  <variable type='float'>x0</variable>
  <variable type='float'>y0</variable>
  <variable type='float'>z0</variable>
</ExternalModel>
```

</Models>

7.4 Code

Description: This is the generic class that import an external code into the framework

Summary

Example

7.5 Projector

Description

Summary

Example

7.6 PostProcessor

Description

List variable, Input Data, Keyword sul tipo analisi statistica!!

Summary

Example

8 Steps

- SingleRun: This is the step that will perform just one evaluation
- MultiRun: This class implements one step of the simulation pattern where several runs are, needed without being adaptive.
 - * name = name of the step (sequence) defined in the RunInfo block, under the Sequence card
 - pauseAtEnd= if True the code will not go to next step until plots are closed manually by the user
 - Sampler:
 - * class=Samplers
 - * type: the type of sampler used in the Samplers block
 - * name of the sampler
 - Model:
 - * class= (Models)
 - * type=(dummy, ROM, External Model, Code, Projector, PostProcessor)
 - * name of the model
 - Input:
 - * class=(Data e Files)
 - * type:
 - if class = files —> none
 - if class = Data —> timepoint, timepointset, historie, histories
 - Output:
 - * class: they are the output destinations
 - Datas
 - OutStreamManager
 - * type:
 - if class=Datas —>(timepoint, timepointset, historie, histories)
 - if class=OutstreamManager —> type:print, plot
 - * name of the output

```

<MultiRun name='***' pauseAtEnd='***'>
  <Sampler class='Samplers' type='***'>***</Sampler>
  <Input class='***' type='***'>***</Input>
  <Model class='Models' type='***'>***</Model>
  <Output class='Datas' *** type='***'</Output>
  <Output class='OutputStreamManager' type='***'>***</Output>
</MultiRun>

```

- Adaptive: this class implement one step of the simulation pattern where several runs are needed in an adaptive scheme

- Sampler: class=samplers type=Adaptive; ???;
- Model: class=Models type=(dummy, ROM, External Model, Code, Projector, PostProcessor)
- Function: it takes in a datas and generate the value of the goal functions, it gives the criteria for which i represent the limit surface (class=Functions)
- Input:
- TargetEvaluation: is the output datas that is used for the evaluation of the goal function, it represents the sampling points. It has to be declared among the outputs.
- SolutionExport: if declared it is used to export the location of the goal functions = 0, it exports the limit surface
- ROM: is boolean, it selects values of input (through sampling) to find the limit surface through the values that were given by the function.
- Output:

```

<Adaptive name='***' pauseAtEnd='***'>
  <Input class = 'Datas' type = 'TimePointSet'>***</Input>
  <Sampler class = 'Samplers' type = 'Adaptive'>***</Sampler>
  <TargetEvaluation class = 'Datas' type = '***'>***</TargetEvaluation>

```

- IODataBase: This step type is used only to extract or push information from/into a DataBase. If in the Databases block the Database is created (no directory or filename) then the Database will be an output of this block, otherwise, if it is uploaded then in this block it will be a Input

- Input: class=(Datas or Databases) type=(if Databases, HDF5, if Datas:timepoint, timepointset, historie, histories)
- Output: class=(Datas or Databases) type=(if Databases, HDF5, if Datas:timepoint, timepointset, historie, histories)
- RomTrainer: This step type is used only to train a ROM.
 - <RomTrainer name='***'>
 - <Input class='Datas' type='TimePointSet'>***</Input>
- PostProcess:
- OutStreamStep:

9 Datas

- TimePoint: A couple of points inside the input and output spaces.
- TimePointSet: A set of couples of points inside the input and output spaces.
- History: A set of couple of points inside the input space and a time dependent array inside the output space. Because time is not a continuous variable inside the RAVEN environment, each array is associated with an array of time points. The input space points and the output space array are correlated by model. (see section *models*)
- Histories: A set of couple of points inside the input space and a set of array inside the output space. The input space points and the output space array are correlated by model. (see section *models*)

Input: sampling variables inside relap or any TH code

Output: variables inside the model output (OutPlaceHolder is a keyword for a something that doesn't have an output)

```
<Datas>
<TimePointSet name='***'>
  <Input>***, ***, ***</Input>
  <Output>***, ***</Output>
</TimePointSet>
</Datas>
```

10 Databases

- name: database name
- directory:
 - if it is not specified a directory, such directory will be created. It will be named DataBaseStorage in the working directory with inside the .h5 file containing the database.
 - if input directory name is given the code will look for DataBaseStorage folder and pick the file from filename from inside such directory
- filename: input file to be read from, if not there such file will be created

Example:

```
<DataBases>  
  <HDF5 name="***" directory='***' filename='***' />  
</DataBases>
```

11 OutStream system

Author: Andrea Alfonsi

The PRA and UQ framework provides the capabilities to visualize and dump out the data that are generated, imported (from a system code) and post-processed during the analysis. These capabilities are contained in the "OutStream" system. Actually, two different OutStream types are available:

- Print, module that lets the user dump the data contained in the internal objects
- Plot, module, based on Matplotlib [?], aimed to provide advanced plotting capabilities

Both the types listed above only accept as "input" a *Data* object type. This choice has been taken since the "*Datas*" system (see section ??) has the main advantages, among the others, of ensuring a standardized approach for exchanging the data/meta-data among the different framework entities. Every module can project its outcomes into a *Data* object. This provides, to the user, the capability to visualize/dump all the modules' results. As already mentioned [put reference to the xml input section], the RAVEN framework input is based on the Extensible Markup Language (XML) format. Thus, in order to activate the "*OutStream*" system, the input needs to contain a block identified by the "< *OutStreamManager* >" tag (as shown below).

```
-----  
<OutStreamManager>  
  <!-- "OutStream" objects that need to be created-->  
</OutStreamManager>  
-----
```

In the "OutStreamManager" block an unlimited number of "Plot" and "Print" sub-blocks can be inputted. The input specifications and the main capabilities for both types are reported in the following sections.

11.1 Printing system

The Printing system has been created in order to let the user dump the data, contained in the internal data objects (see [reference to Data(s) section]), out at anytime during the calculation. Currently, the only available output is a Comma Separated Value (CSV) file.

In the near future, an XML formatted file option will be available. This will facilitate the exchanging of results and provide the possibility to dump the solution of an analysis and "restart" another one constructing a *Data* from scratch. The XML code, that is reported below, shows different ways to request a *Print* OutputStream. The user needs to provide a name for each sub-block (XML attribute). These names are then used in the *Steps*' blocks in order to activate the Printing options at anytime. As shown in the examples below, every *Print* block must contain, at least, the two required tags:

- `< type >`, the output file type (csv or xml). *Note, only csv is currently available*
- `< source >`, the *Data* name (one of the *Data* defined in the "Datas" block)

If only these two tags are provided (as in the "first-example" below), the output file will be filled with the whole content of the "d-name" *Data*.

```
-----
<OutputStreamManager>
  <Print name='first_example'>
    <type>csv</type>
    <source>d-name</source>
  </Print>
  <Print name='second-example'>
    <type>csv</type>
    <source>d-name</source>
    <variables>Output</variables>
  </Print>
  <Print name='third-example'>
    <type>csv</type>
    <source>d-name</source>
    <variables>Input</variables>
  </Print>
  <Print name='forth-example'>
    <type>csv</type>
    <source>d-name</source>
    <variables>Input|var-name-in,Output|var-name-out</variables>
  </Print>
</OutputStreamManager>
-----
```

If just few parts of the `< source >` are important for a particular analysis, the additional XML tag `< variables >` can be provided. In this block, the variables that need to be

dumped must be inputted, in a comma separated format. The available options, for the *< variables >* sub-block, are listed below:

- **Output**, the output space will be dumped out (see “second-example”)
- **Input**, the input space will be dumped out (see “third-example”)
- **Input—var-name-in/Output—var-name-out**, only the particular variables “var-name-in” and “var-name-out” will be reported in the output file (see “forth-example”)

Note that all the XML tags are case-sensitive but not their content.

11.2 Plotting system

The Plotting system provides all the capabilities to visualize the analysis outcomes, in real-time or at the post-processing stage. The system is based on the Python library Matplotlib [?]. Matplotlib is a 2D/3D plotting library which produces publication quality figures in a variety of hardcopy formats and interactive environments across platforms. This external tool has been wrapped in the RAVEN framework, and is usable by the user. Since it was unfeasible to support, in the source code, all the interfaces for all the available plot types, the RAVEN Plotting system directly provide a formatted input structure for 11 different plot types (2D/3D). The user may request a plot not present among the supported ones, since the RAVEN Plotting system has the capability to construct on the fly the interface for a Plot, based on XML instructions. This capability will be discussed in the sub-section ??.

11.2.1 Plot input structure

In order to create a plot, the user needs to add, within the *< OutStreamManager >* block, a *< Plot >* sub-block. As for the *Print OutStream*, the user needs to specify a name as attribute of the plot. This name will then be used to request the plot in the *Steps'* block. In addition, the Plot block may need the following attributes:

- **dim**, *required integer attribute*, define the dimensionality of the plot: 2 (2D) or 3 (3D)
- **interactive**, *optional bool attribute (default=False)'*, specify if the Plot needs to be interactively created (real-time screen visualization)

- **overwrite**, *optional bool attribute (default=False)*’, if the plot needs to be dumped into picture file/s, does the code need to overwrite them every time a new plot (with the same name) is requested?

As shown, in the XML input example below, the body of the Plot XML input contains two main sub-nodes:

- *< actions >*, where general control options for the figure layout are defined (see [])
- *< plot_settings >*, where the actual plot options are provided

These two main sub-block are discussed in the following paragraphs.

11.2.1.1 “Actions” input block

The input in the *< actions >* sub-node is common to all the Plot types, since, in it, the user specifies all the controls that need to be applied to the figure style. This block must be unique in the definition of the *< Plot >* main block. In the following list, all the predefined “actions” are reported:

- *< how >*, comma separated list of output types:
 - *screen*, show the figure on the screen in interactive mode
 - *pdf*, save the figure as a Portable Document Format file (PDF)
 - *png*, save the figure as a Portable Network Graphics file (PNG)
 - *eps*, save the figure as a Encapsulated Postscript file (EPS)
 - *pgf*, save the figure as a LaTeX PGF Figure file (PGF)
 - *ps*, save the figure as a Postscript file (PS)
 - *gif*, save the figure as a Graphics Interchange Format (GIF)
 - *svg*, save the figure as a Scalable Vector Graphics file (SVG)
 - *jpeg*, save the figure as a jpeg file (JPEG)
 - *raw*, save the figure as a Raw RGBA bitmap file (RAW)
 - *bmp*, save the figure as a Windows bitmap file (BMP)
 - *tiff*, save the figure as a Tagged Image Format file (TIFF)
 - *svgz*, save the figure as a Scalable Vector Graphics file (SVGZ)

- *< title >*, as the name suggests , within this block the user can specify the title of the figure. In the body, few other keywords (required and not) are present:
 - *< text >*, string type, title of the figure
 - *< kwargs >*, within this block the user can specify optional parameters with the following format:

```

-----
<kwargs>
  <param1>value1</param1>
  <param2>value2</param2>
</kwargs>
-----

```

The kwargs block is able to convert whatever string into a python type (for example *< param1 > '1stKeyword' : 45 < /param1 >* will be converted into a dictionary, *< param2 > [56,67] < /param2 >* into a list, etc.). For reference regarding the available kwargs, see “matplotlib.pyplot.title” method in [?].

- *< label_format >*, within this block the default scale formatting can be modified. In the body, few keywords can be specified (all optional):
 - *< style >*, string, the style of the number notation, 'sci' or 'scientific' for scientific, 'plain' for plain notation. Default = scientific
 - *< scilimits >*, tuple, (m, n), pair of integers; if style is 'sci', scientific notation will be used for numbers outside the range 10^m to 10^n . Use (0,0) to include all numbers. NB. The value for this keyword, needs to be inputted between brackets [for example, (5,6)]. Default = (0,0)
 - *< useOffset >*, bool or double, if True, the offset will be calculated as needed; if False, no offset will be used; if a numeric offset is specified, it will be used. Default = False
 - *< axis >*, string, the axis where to apply the defined format, 'x', 'y' or 'both'. Default = 'both'. NB. If this action will be used in a 3-D plot, the user can input 'z' as well and 'both' will apply this format to all three axis.
- *< figure_properties >*, within this block the user specifies how to customize the figure style/quality. Thus, through this “action” the user has got full control on the quality of the figure, its dimensions, etc. This control is performed by the following keywords:

- *< figsize >*, tuple (optional), (width, height), in inches
- *< dpi >*, integer, dots per inch
- *< facecolor >*, string, set the figure background color (please refer to “matplotlib.figure.Figure” in [?] for a list of all the colors available)
- *< edgecolor >*, string, the figure edge background color (please refer to “matplotlib.figure.Figure” in [?] for a list of all the colors available)
- *< linewidth >*, self explainable keyword
- *< frameon >*, bool, if False, suppress drawing the figure frame
- *< range >*, the range “action” allows to specify the ranges of all the axis. All the keywords in the body of this block are optional:
 - *< ymin >*, double (optional), lower boundary for y axis
 - *< ymax >*, double (optional), upper boundary for y axis
 - *< xmin >*, double (optional), lower boundary for x axis
 - *< xmax >*, double (optional), upper boundary for x axis
 - *< zmin >*, double (optional), lower boundary for z axis. NB. Obviously, this keyword is effective in 3-D plots only
 - *< zmax >*, double (optional), upper boundary for z axis. NB. Obviously, this keyword is effective in 3-D plots only
- *< camera >*, the camera item is available in 3-D plots only. Through this “action”, it is possible to orientate the plot as wished. The controls are:
 - *< elevation >*, double (optional), stores the elevation angle in the z plane
 - *< azimuth >*, double (optional), stores the azimuth angle in the x,y plane
- *< scale >*, the scale block allows the specification of the axis scales:
 - *< xscale >*, string (optional), scale of the x axis. Three options are available: “linear”, “log”, “symlog”. Default = linear
 - *< yscale >*, string (optional), scale of the y axis. Three options are available: “linear”, “log”, “symlog”. Default = linear
 - *< zscale >*, string (optional), scale of the z axis. Three options are available: “linear”, “log”, “symlog”. Default = linear. NB. Obviously, this keyword is effective in 3-D plots only

- `< add_text >`, same as title
- `< autoscale >`, the autoscale block is a convenience method for simple axis view autoscaling. It turns autoscaling on or off, and then, if autoscaling for either axis is on, it performs the autoscaling on the specified axis or axes. The following keywords are available:
 - `< enable >`, bool (optional), True turns autoscaling on, False turns it off. None leaves the autoscaling state unchanged. Default = True
 - `< axis >`, string (optional), string, the axis where to apply the defined format, 'x', 'y' or 'both'. Default = 'both'. NB. If this action will be used in a 3-D plot, the user can input 'z' as well and 'both' will apply this format to all three axis.
 - `< tight >`, bool (optional), if True, set view limits to data limits; if False, let the locator and margins expand the view limits; if None, use tight scaling if the only artist is an image, otherwise treat tight as False.
- `< horizontal_line >`, this “action” provides the ability to draw a horizontal line in the current figure. This capability might be useful, for example, if the user wants to highlight a trigger, function of a variable. The following keywords are settable:
 - `< y >`, double (optional), y coordinate. Default = 0
 - `< xmin >`, double (optional), starting coordinate on the x axis. Default = 0
 - `< xmax >`, double (optional), ending coordinate on the x axis. Default = 1
 - `< kwargs >`, within this block the user can specify optional parameters with the following format:

```

-----
<kwargs>
  <param1>value1</param1>
  <param2>value2</param2>
</kwargs>
-----

```

The kwargs block is able to convert whatever string into a python type (for example `< param1 > '1stKeyword' : 45 < /param1 >` will be converted into a dictionary, `< param2 > [56,67] < /param2 >` into a list, etc.). For reference regarding the available kwargs, see “matplotlib.pyplot.axhline” method in [?].

NB. This capability is not available for 3-D plots.

- `< vertical_line >`, similarly to the “horizontal_line” action, this block provides the ability to draw a vertical line in the current figure. This capability might be useful, for example, if the user wants to highlight a trigger, function of a variable. The following keywords are settable:

- `< x >`, double (optional), x coordinate. Default = 0
- `< ymin >`, double (optional), starting coordinate on the y axis. Default = 0
- `< ymax >`, double (optional), ending coordinate on the y axis. Default = 1
- `< kwargs >`, within this block the user can specify optional parameters with the following format:

```

-----
<kwargs>
  <param1>value1</param1>
  <param2>value2</param2>
</kwargs>
-----

```

The kwargs block is able to convert whatever string into a python type (for example `< param1 > '1stKeyword' : 45 < /param1 >` will be converted into a dictionary, `< param2 > [56,67] < /param2 >` into a list, etc.). For reference regarding the available kwargs, see “matplotlib.pyplot.axvline” method in [?].

NB. This capability is not available for 3-D plots.

- `< horizontal_rectangle >`, this “action” provides the possibility to draw, in the current figure, a horizontally orientated rectangle . This capability might be useful, for example, if the user wants to highlight a zone in the plot. The following keywords are settable:
- `< ymin >`, double (required), starting coordinate on the y axis
 - `< ymax >`, double (required), ending coordinate on the y axis
 - `< xmin >`, double (optional), starting coordinate on the x axis. Default = 0
 - `< xmax >`, double (optional), ending coordinate on the x axis. Default = 1
 - `< kwargs >`, within this block the user can specify optional parameters with the following format:

```

-----
<kwargs>
  <param1>value1</param1>
  <param2>value2</param2>
</kwargs>
-----

```

The kwargs block is able to convert whatever string into a python type (for example `< param1 > '1stKeyword' : 45 < /param1 >` will be converted into a dictionary, `< param2 > [56,67] < /param2 >` into a list, etc.). For reference regarding the available kwargs, see “matplotlib.pyplot.axhspan” method in [?].

NB. This capability is not available for 3-D plots.

- `< vertical_rectangle >`, this “action” provides the possibility to draw, in the current figure, a vertically orientated rectangle . This capability might be useful, for example, if the user wants to highlight a zone in the plot. The following keywords are settable:
 - `< xmin >`, double (required), starting coordinate on the x axis
 - `< xmax >`, double (required), ending coordinate on the x axis
 - `< ymin >`, double (optional), starting coordinate on the y axis. Default = 0
 - `< ymax >`, double (optional), ending coordinate on the y axis. Default = 1
 - `< kwargs >`, within this block the user can specify optional parameters with the following format:

```

-----
<kwargs>
  <param1>value1</param1>
  <param2>value2</param2>
</kwargs>
-----

```

The kwargs block is able to convert whatever string into a python type (for example `< param1 > '1stKeyword' : 45 < /param1 >` will be converted into a dictionary, `< param2 > [56,67] < /param2 >` into a list, etc.). For reference regarding the available kwargs, see “matplotlib.pyplot.axvspan” method in [?].

NB. This capability is not available for 3-D plots.

- `< axes_box >`, this keyword controls the axes' box. No body and its value can be 'on' or 'off'.
- `< axis_properties >`, this block is used to set axis properties. There are not fixed keywords. If only a single property needs to be set, it can be specified as body of this block, otherwise a dictionary-like string needs to be provided. For reference regarding the available keys, refer to “matplotlib.pyplot.axis” method in [?].
- `< grid >`, this block is used to define a grid that needs to be added in the plot. The following keywords can be inputted:
 - `< b >`, double (required), starting coordinate on the x axis
 - `< which >`, double (required), ending coordinate on the x axis
 - `< axis >`, double (optional), starting coordinate on the y axis. Default = 0
 - `< kwargs >`, within this block the user can specify optional parameters with the following format:

```
-----  
<kwargs>  
  <param1>value1</param1>  
  <param2>value2</param2>  
</kwargs>  
-----
```

The kwargs block is able to convert whatever string into a python type (for example `< param1 > '1stKeyword' : 45 < /param1 >` will be converted into a dictionary, `< param2 > [56,67] < /param2 >` into a list, etc.). For reference regarding the available kwargs, see “matplotlib.pyplot.grid” method in [?].

11.2.1.2 “plot_settings” input block

The sub-block identified by the keyword `< plot_settings >` is used to define the plot characteristics. Within this sub-section at least a `< plot >` block must be present. the `< plot >` sub-section may not be unique within the `< plot_settings >` definition; the number of `< plot >` sub-blocks is equal to the number of plots that need to be placed in the same figure. For example, in the following XML cut, a “line” and a “scatter” type are combined in the same figure.

```

<OutputStreamManager>
  <Plot name='example2PlotsCombined' dim='2'>
    <actions>
      <!-- Actions -->
    </actions>
    <plot\_settings>
      <plot>
        <type>line</type>
        <x>d-type|Output|x1</x>
        <y>d-type|Output|y1</y>
      </plot>
      <plot>
        <type>scatter</type>
        <x>d-type|Output|x2</x>
        <y>d-type|Output|y2</y>
      </plot>
      <xlabel>label X</xlabel>
      <ylabel>label Y</ylabel>
    </plot\_settings>
  </Plot>
</OutputStreamManager>

```

As already mentioned, within the `< plot_settings >` block, at least a `< plot >` sub-block needs to be inputted. Independently from the plot type, some keywords are mandatory:

- `< type >`, string, the plot type (for example, line, scatter, wireframe, etc.)
- `< x >`, string, the parameter needs to be considered as x coordinate
- `< y >`, string, the parameter needs to be considered as y coordinate
- `< z >`, string (required in 3-D plots only), the parameter needs to be considered as z coordinate

In addition other plot-dependent keywords, reported in section ??, can be provided. Under the `< plot_settings >` block other keywords, common to all the plots the user decided to combine in the figure, can be inputted, such as:

- `< xlabel >`, string, x axis label
- `< ylabel >`, string, y axis label
- `< zlabel >`, string (available in 3-D plots only), z axis label

11.2.1.3 Predefined Plotting System: 2D/3D

sgagagga _____

`<OutputStreamManager>`

`<Plot name='2DHistoryPlot' dim='2' interactive=False' overwrite=False'>`

`<actions>`

`<how>pdf,png,eps</how>`

`<title>`

`<text> </text>`

`</title>`

`</actions>`

`<plot_settings>`

`<plot>`

`<type>line</type>`

`<x>stories|Output|time</x>`

`<y>stories|Output|pipe1_Hw</y>`

`<kwargs>`

`<color>green</color>`

`<label>pipe1-Hw</label>`

`</kwargs>`

`</plot>`

`<plot>`

`<type>line</type>`

`<x>stories|Output|time</x>`

`<y>stories|Output|pipe1_aw</y>`

`<kwargs>`

`<color>blue</color>`

`<label>pipe1-aw</label>`

`</kwargs>`

`</plot>`

`<xlabel>time [s]</xlabel>`

```
<ylabel>evolution</ylabel>  
</plot_settings>  
</Plot>  
</OutputStreamManager>
```

11.2.2 Interpreted Plotting instruction

12 Existing Interfaces

12.1 RELAP5 Interface

12.1.1 Files

In the `< Files >` section, as specified before, there should be specified all the files needed for the code to run. In the case of RELAP5 most of the times the files needed are the following:

- RELAP5 Input file
- Table file or files that RELAP needs to run
- relapdata.py
- relap5run.py

Example:

```
<Files>X10.i,tpfh2o,relapdata.py,RELAP5run.py</Files>
```

It is a good practice to put inside the working directory all of these files and also:

- the RAVEN input file
- the executable file for RELAP5
- the license for the executable for RELAP5

12.1.2 Sequence

In the `< Sequence >` section there should be specified the names of the steps declared in the `< Steps >` block. So if we called the first multirun: "Grid_Sampler" and the second multirun: "MC_Sampler" in the sequence section we should see this:

```
<Sequence>Grid_Sampler,MC_Sampler</Sequence>
```

12.1.3 batchSize and mode

For the `< batchSize >` and `< mode >` sections please take a look at the RunInfo block in the previous chapters.

12.1.4 RunInfo

After all of these blocks are filled out here is a standard example of what a RunInfo block can look like:

```
<RunInfo>
  <WorkingDir>~/working_dir</WorkingDir>
  <Files>inputfilerelap.i,tpfh2o,relapdata.py,RELAP5run.py</Files>
  <Sequence>Grid_Sampler,MC_Sampler</Sequence>
  <batchSize>1</batchSize>
    <mode>pbsdsh</mode>
  <expectedTime>1:00:00</expectedTime>
  <ParallelProcNumb>1</ParallelProcNumb>
</RunInfo>
```

12.1.5 Models

For the `< Models >` block here is a standard example of what can be used to use RELAP5 as the external model:

```
<Models>
  <Code name='MyRELAP' subType='Relap5'><executable>python
    RELAP5run.py</executable></Code>
</Models>
```

12.1.6 Distributions

In the `< Distribution >` block are defined the distributions that are going to be used for the sampling of the variables defined in the `< Samplers >` block. For all the possible

distributions and all their possible inputs please see the chapter about Distributions. Here we give a general example of three different distributions, with the names:

```
<Distributions debug='True'>
  <Triangular name='BPfailtime'>
    <apex>5.0</apex>
    <min>4.0</min>
    <max>6.0</max>
  </Triangular>
  <LogNormal name='BPrepairtime'>
    <mean>0.75</mean>
    <sigma>0.25</sigma>
  </LogNormal>
  <Uniform name='ScalFactPower'>
    <low>1.0</low>
    <hi>1.2</hi>
  </Uniform>
</Distributions>
```

It is good practice to name the distribution something similar to what kind of variable is going to be sampled, since there might be many variables with the same kind of distributions but different input parameters.

12.1.7 Samplers

In the `< Samplers >` block we want to define the variables that are going to be sampled.

Example: We want to do the sampling of 3 variables:

- Battery Fail Time
- Battery Repair Time
- Scaling Factor Power Rate

We are going to sample these 3 variables using two different sampling methods: grid and MonteCarlo.

In RELAP5 the sampler reads the variable in this way: Given the name, the first number is the card number, the second number is the word number. In this example we are sampling:

- For card 0000588 (trip) the word 6 (battery failure time)
- For card 0000575 (trip) the word 6 (battery repair time)
- For card 20210000 (reactor power) the word 4 (reactor scaling factor)

We proceed to do so for both the Grid sampling and the MonteCarlo sampling.

```
<Samplers debug='True'>
  <Grid name='Grid_Sampler' initial_seed='210491' >
    <variable name='0000588:6'>
      <distribution>BPfailtime</distribution>
      <grid type='value' construction='equal' lowerBound='0.0'
        steps='10'>2880</grid>
    </variable>
    <variable name='0000575:6'>
      <distribution>BPrepairtime</distribution>
      <grid type='value' construction='equal' lowerBound='0.0'
        steps='10'>2880</grid>
    </variable>
    <variable name='20210000:4'>
      <distribution>ScalFactPower</distribution>
      <grid type='value' construction='equal' lowerBound='1.0'
        steps='10'>0.02</grid>
    </variable>
  </Grid>
  <MonteCarlo name='MC_Sampler' limit='1000'>
    <variable name='0000588:6'>
      <distribution>BPfailtime</distribution>
    </variable>
    <variable name='0000575:6'>
      <distribution>BPrepairtime</distribution>
    </variable>
    <variable name='20210000:4'>
      <distribution>ScalFactPower</distribution>
    </variable>
  </MonteCarlo>
</Samplers>
```


</MonteCarlo>
</Samplers>

It can be seen that each variable is connected with a proper distribution defined in the < *Distributions* > block. Here is how the Input would be read for the first variable:

We are sampling a variable situated in word 6 of the card 0000588 using a Grid sampling method. The distribution that this variable is following is a Triangular distribution (see section above). We are sampling this variable beginning from 0.0 in 10 *equal* steps of 2880.

12.1.8 Steps

For a RELAP interface the steps that are probably going to be used are the < *MultiRun* >. First we need to name the step: this name is one of the ones used in the < *Sequence* > block. In this case: Grid_Sampler and MC_Sampler.

<MultiRun name='Grid_Sampler' debug='True' re-seeding='210491'>

With this step we need to import all the files needed for the simulation:

- RELAP input file
- RELAP5run.py
- relapdata.py
- elements tables – tpfh2o

<Input class='Files' type=''>inputrelap.i</Input>
<Input class='Files' type=''>RELAP5run.py</Input>
<Input class='Files' type=''>relapdata.py</Input>
<Input class='Files' type=''>tpfh2o</Input>

We then need to define which model it is used:

<Model class='Models' type='Code'>MyRELAP</Model>

We then need to specify which Sampler is used, and this can be done as follows:

```
<Sampler class='Samplers' type='Grid'>Grid_Sampler</Sampler>
```

And lastly we need to specify what kind of Output the user wants. For example the user might want to make a DataBase (in RAVEN the DataBase created is a HDF5 file. Here is a classical example:

```
<Output class='DataBases' type='HDF5'>MC_out</Output>
```

Following is the example of two MultiRun steps which use different sampling methods (grid and montecarlo), and creating two different DataBases for each one:

```
<Steps debug='True'>
  <MultiRun name='Grid_Sampler' debug='True' re-seeding='210491'>
    <Input class='Files' type=''>X10.i</Input>
    <Input class='Files' type=''>RELAP5run.py</Input>
    <Input class='Files' type=''>relapdata.py</Input>
    <Input class='Files' type=''>tpfh2o</Input>
    <Model class='Models' type='Code'>MyRELAP</Model>
    <Sampler class='Samplers' type='Grid'>Grid_Sampler</Sampler>
    <Output class='DataBases' type='HDF5'>Grid_out</Output>
  </MultiRun>
  <MultiRun name='MC_Sampler' debug='True' re-seeding='210491'>
    <Input class='Files' type=''>X10.i</Input>
    <Input class='Files' type=''>RELAP5run.py</Input>
    <Input class='Files' type=''>relapdata.py</Input>
    <Input class='Files' type=''>tpfh2o</Input>
    <Model class='Models' type='Code'>MyRELAP</Model>
    <Sampler class='Samplers' type='MonteCarlo'>MC_Sampler</Sampler>
    <Output class='DataBases' type='HDF5'>MC_out</Output>
  </MultiRun>
</Steps>
```

12.1.9 DataBases

As shown in the `< Steps >` block, the code is creating two DataBases called Grid_out and MC_out. So the user needs to input the following

```
<DataBases>
```

```

    <HDF5 name="Grid_out"/>
    <HDF5 name="MC_out"/>
</DataBases>

```

As listed before, this will create two DataBases. The files are .h5 extension files.

12.2 RELAP7 Interface

12.2.1 Files

In the `< Files >` section, as specified before, there should be specified all the files needed for the code to run. In the case of RELAP7 most of the times the files needed are the following:

- RELAP7 Input file
- Control Logic file

Example:

```

<Files>nat_circ, control_logic.py</Files>

```

12.2.2 Models

For the `< Models >` block RELAP7 uses the RAVEN executable. Here is a standard example of what can be used to use RELAP7 as the model:

```

<Models>
  <Code name='MyRAVEN'
    subType='RAVEN'><executable>~path/to/RAVEN-opt</executable></Code>
</Models>

```

12.2.3 Distributions

The `< Distributions >` block, when using RELAP7 has to be specified also through the control logic. Given the names of the distributions, and their parameters, a python file should be used for the control logic. For example, it is required the sampling of a normal distribution for the primary pressure in RELAP7.

```
<Distributions>
  <Normal name="Prim_Pres">
    <mean>1000000</mean>
    <sigma>100</sigma>
  </Normal>
</Distributions>
```

The python file associated to it should look like this:

```
def initial_function(monitored , controlled , auxiliary )
    print("monitored",monitored,"controlled",
        controlled,"auxiliary",auxiliary)

    controlled.pressure_in_pressurizer =
        distributions.Prim_Pres.getDistributionRandom()
    return
```

12.2.4 Samplers

In the `< Samplers >` block there are going to be defined both which kind of sampling method used, the variables not sampled inside the control logic and of course which distributions will follow the chosen sampling method. For the example it was chosen a Monte-Carlo sampling with a 500 runs. The global initial pressure wasn't specified in the control logic so it is sampled in this block. It is also specified that such initial pressure has to follow the same distribution as the primary pressure.

```
<Samplers>
  <MonteCarlo name="MC_samp" limit="500">
    <variable name="GlobalParams|global_init_P">
      <distribution>Prim_Pres</distribution>
    </variable>
```

</MonteCarlo>
</Samplers>

12.2.5 Steps

