

Methodology of Uncertainty Management

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Examples with OpenTURNS and Uranie'
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MAISON DE LA SIMULATION



CHANGER L'ÉNERGIE ENSEMBLE

Which uncertainty sources?

- ▶ The modeling process of a phenomenon contains many sources of uncertainty:
 - model uncertainty: the translation of the phenomenon into a set of equations. The understanding of the physicist is always incomplete and simplified,
 - numerical uncertainty: the resolution of this set of equations often requires some additional numerical simplifications,
 - parametric uncertainty: the user feeds in the model with a set of deterministic values ... According to his/her knowledge
- ▶ Different kinds of uncertainties taint engineering studies; **we focus here on parametric uncertainties** (as it is common in practice)

Which (parametric) uncertainty sources?

► Epistemic uncertainty

- It is related to the **lack of knowledge** or precision of any given parameter which is deterministic in itself (or which could be considered as deterministic under some accepted hypotheses). E.g. a characteristic of a material.

► Stochastic (or aleatory) uncertainty

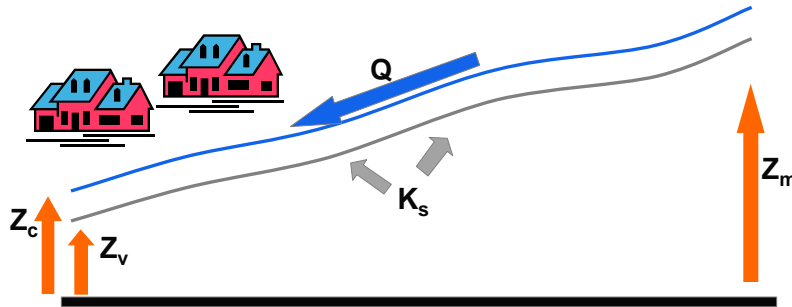
- It is related to the **real variability of a parameter**, which cannot be reduced (e.g. the discharge of a river in a flood risk evaluation). The parameter is stochastic in itself.

► Reducible vs non-reducible uncertainties

- Epistemic uncertainties are (at least theoretically) reducible
- Instead, stochastic uncertainties are (in general) irreducible (the discharge of a river will never be predicted with certainty)
 - A counter-example: stochastic uncertainty tainting the geometry of a mechanical piece → Can be reduced by improving the manufacturing line ... **The reducible aspect is quite relative** since it depends on whether the cost of the reduction actions is affordable in practice

A (very) simplified example

Flood water level calculation



Uncertainty

$$Z_c = Z_v + \left[\frac{Q}{K_s \cdot \sqrt{(Z_m - Z_v)/L} \cdot B} \right]^{3/5}$$

Strickler's Formula

- ◆ Z_c : Flood level (variable of interest)
- ◆ Z_m et Z_v : level of the riverbed, upstream and downstream (random)
- ◆ Q : river discharge (random)
- ◆ K_s : Strickler's roughness coefficient (random)
- ◆ B, L : Width and length of the river cross section (deterministic)

General framework

Input Variables

Uncertain : X
Fixed : d

Model

$G(X, d)$

Output variables of interest

$Z = G(X, d)$

Which output variable of interest?

- Formally, we can link the output variable of interest Z to a number of continuous or discrete uncertain inputs X through the function G :

$$Z = G(X, d)$$

- d denotes the “fixed” variables of the study, representing, for instance a given scenario. In the following we will simply note:

$$Z = G(X)$$

- The dimension of the output variable of interest can be 1 or >1
- Function G can be presented as:
 - an analytical formula or a complex finite element code,
 - with high / low computational costs (measured by its CPU time),
- The uncertain inputs are modeled thanks to a **random vector** X , composed of n univariate random variables (X_1, X_2, \dots, X_n) linked by a dependence structure.

Which goal?

► Four categories of industrial objectives:

- Industrial practice shows that the goals of any quantitative uncertainty assessment usually fall into the following four categories:
 - **Understanding**: to understand the influence or rank importance of uncertainties, thereby guiding any additional measurement, modeling or R&D efforts.
 - **Accrediting**: to give credit to a model or a method of measurement, i.e. to reach an acceptable quality level for its use.
 - **Selecting**: to compare relative performance and optimize the choice of a maintenance policy, an operation or design of the system.
 - **Complying**: to demonstrate the system's compliance with an explicit criteria or regulatory threshold (e.g. nuclear or environmental licensing, aeronautical certification, ...)
- There may be several goals in any given study or along the time: for instance, importance ranking may serve as a first study in a more complex and long study leading to the final design and/or the compliance demonstration

Which criteria?

► Different quantities of interest

- These different objectives are embodied by different criteria upon the output variable of interest.

► These criteria can focus on the outputs':

- range
- central dispersion
- “central” value: mean, median
- probability of exceeding a threshold : usually, the threshold is extreme. For example, in the certification stage of a product.

► Formally, the quantity of interest is a particular feature of the pdf of the variable of interest Z

The major steps in an uncertainty study

► Step A : Problem's specification

- Input variable(s) - Model - Variable(s) of interest – Quantities of interest

► Step B : Quantification of uncertainty sources

► Step C : Propagation of uncertainty sources from inputs to outputs

► Step C' : Sensitivity analysis, Ranking uncertainty sources

More generic steps
(statistics & probability)

Step A : specific to the
problem to be solved

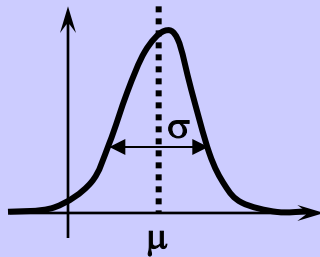
Why are these questions so important?

► The proper identification of:

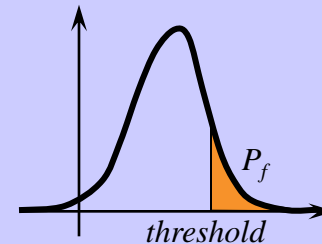
- the uncertain input parameters and the nature of their uncertainty sources,
- the output variable of interest and the goals of a given uncertainty assessment,

► is the key step in the uncertainty study, as it guides the choice of the most relevant mathematical methods to be applied

What is **really** relevant in the uncertainty study?



Mean, median, variance,
(moments) of Z



(Extreme) quantiles, probability of
exceeding a given threshold

A particular quantity of interest: the “probability of failure”

► G models a system (or a part of it) in operative conditions

- Variable of interest $Z \rightarrow$ a given state-variable of the system (e.g. a temperature, a deformation, a water level etc.)

► Following an “operator’s” point of view

- The system is in safe operating condition if Z is above (or below) a given “safety” threshold $Z \leq 0$

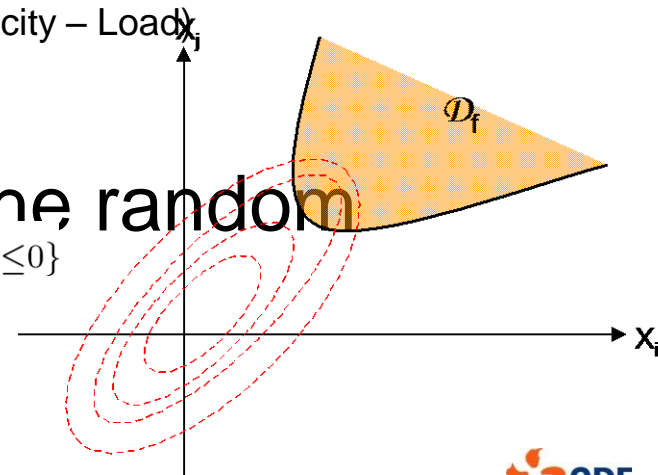
► System “failure” event:

- Classical formulation (no loss of generality) in which the threshold is 0 and the system fails when Z is negative
- Structural Reliability Analysis (SRA) “vision”: Failure if $C-L \leq 0$ (Capacity – Load)

► Failure domain: $\mathcal{D}_f = \{x \in \mathcal{X} : G(x) \leq 0\}$

► Problem: estimating the mean of the random variable “failure indicator”:

$$p_f = \int_{\mathcal{D}_f} f(x) dx = \int_{\mathcal{X}} I_{\mathcal{D}_f}(x) f(x) dx = \mathbb{E} [I_{\mathcal{D}_f}(X)]$$



Need of a generic and shared methodology

- ▶ There has been a considerable rise of interest in many industries in the recent decade
- ▶ Facing the questioning of their control authorities in an increasing number of different domains or businesses, large industrial companies have felt that **domain-specific approaches are no more appropriate**.
- ▶ In spite of the diversity of terminologies, **most of these methods share in fact many common algorithms**.
- ▶ That is why many industrial companies and public establishments have set up **a common methodological framework** which is **generic** to all industrial branches. This methodology has been drafted from industrial practice, which enhances its adoption by industries.

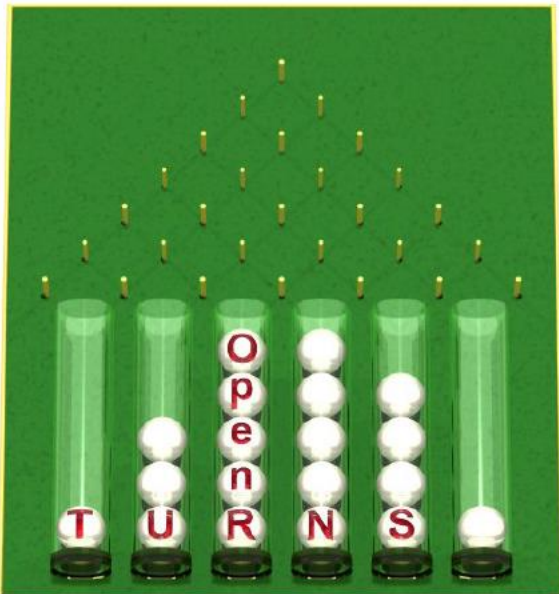
Open TURNS Reference Guide

Reference Guide

OpenTURNS 1.3

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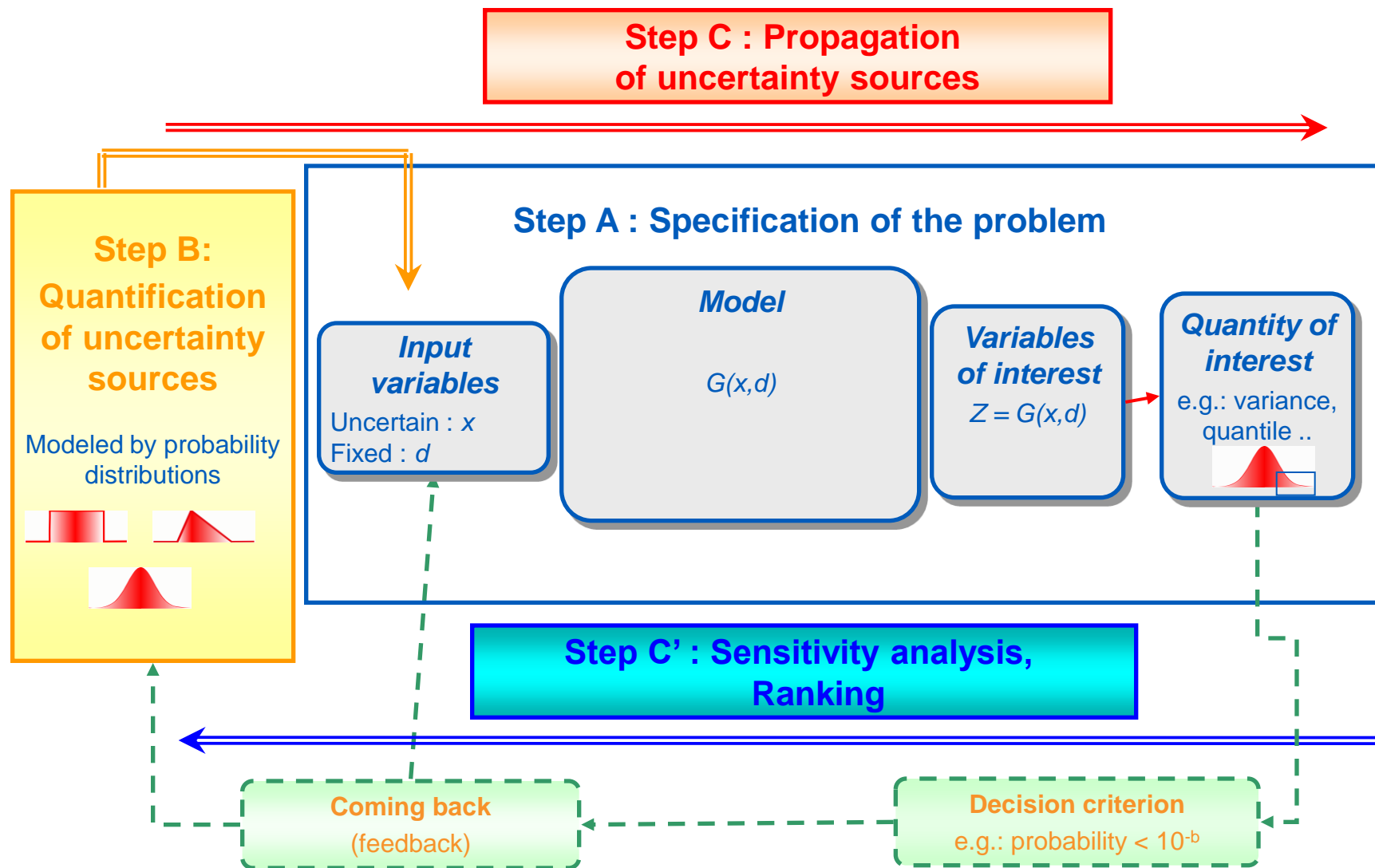


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Uncertainty management - the global methodology



Propagating uncertainties: a tricky problem

► Uncertainty analysis leads to the estimation of some particular features of the output Y

■ An important example: the *probability of failure*

- A system is in safe operating condition if Y is above (or below) a given “safety” threshold. Let us say if $Z > 0$.

$$\mathcal{D}_f = \{x \in \mathcal{X} : G(x) \leq 0\}$$
$$p_f = \int_{\mathcal{D}_f} f(x) dx = \int_{\mathcal{X}} I_{\mathcal{D}_f}(x) f(x) dx = \mathbb{E} [I_{\mathcal{D}_f}(X)]$$

In practical application the order of magnitude of p_f could be 10^{-3} down to 10^{-7}

► Most of time, sampling-based

- Brute force Monte Carlo methods are inappropriate: slow convergence leads to an unaffordable computational burden

► Different (and complementary) strategies:

- Adopting a sampling strategy which reduces the variance of the estimator of the desired quantity
- Building a surrogate model (“meta-model”) the computational cost of which is null, then propagating uncertainty through the meta-model
 - The computational budget is here used to build the meta-model, following a particular DOE

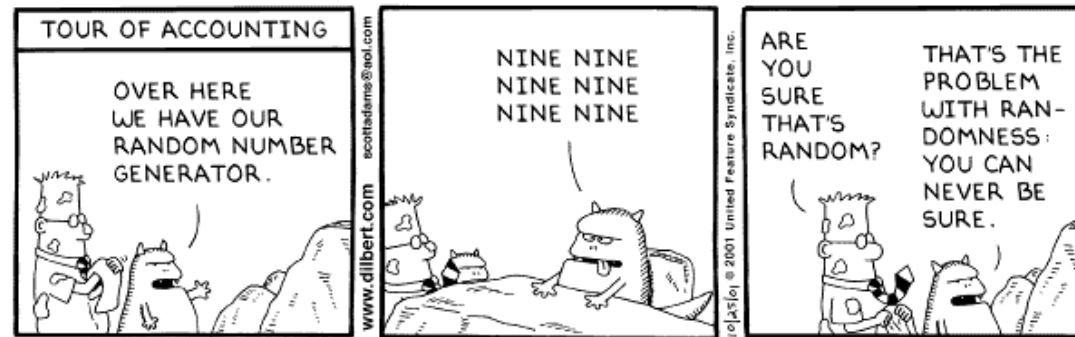
Propagating uncertainties: take-home messages

◆ Main issue in practice with complex models: **the computational burden!**

◆ Avoid DIY solutions!

◆ You may need:

- Appropriate math. methods
(advanced sampling, meta-modeling)
- Appropriate software tools for:
 - Effectively linking the deterministic model $G(X)$ and the probabilistic model of X
 - Performing distributing computations
(High Performance Computing)

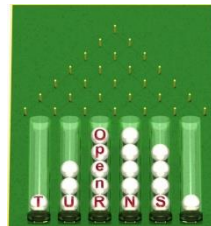
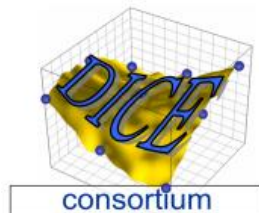


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The “Uncertainty Analysis Community”

- ◆ Several collaborative frameworks at the interface industry-academics
 - multi-partners projects, funded by public powers
 - Industrial partnerships for developing Open Source Software (e.g. OpenTURNS)
 - Research consortiums
 - Working groups within scientific societies



ESReDA
European Safety, Reliability & Data Association



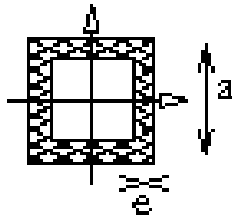
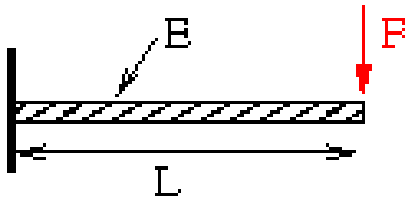
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It's up to you with OpenTURNS and Uranie

► Train yourself on a simple example:

- deviation of a beam on which we apply a load when its other extremity is fixed in a wall
- we model uncertainties on the parameters E , F , L , I
- we apply some of the algorithms of the Methodology using OpenTURNS or Uranie



$$Y = \frac{FL^3}{3EI}$$

- statistical fitting
- dependency modeling
- mean and variance estimation
- sensitivity analysis
- rare event probability estimation using Monte Carlo and FORM estimators
- ...

Consider a diving board in a swimming pool where many kids are jumping....

What about the deviation of the diving board ... and the safety of people just under??

Thank you for your attention

