



DE LA RECHERCHE À L'INDUSTRIE

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Multi-fidelity modelling strategy based on a system approach including NURBS metamodels

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Introduction

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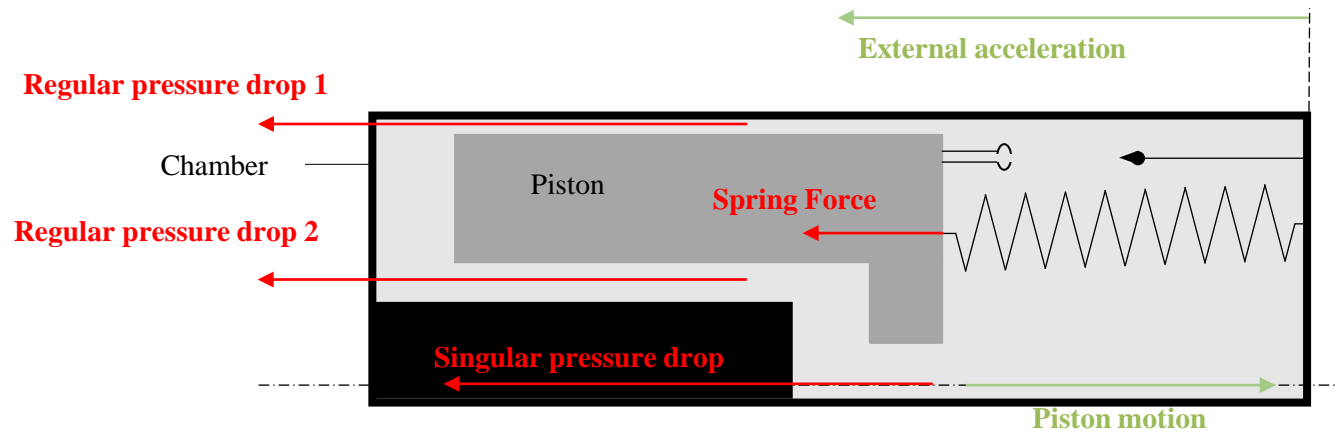
Context

- Difficulties in conducting experiments (more complex experiences);
- Increasing demand for computer resources;
- Emergence of metamodeling strategies.

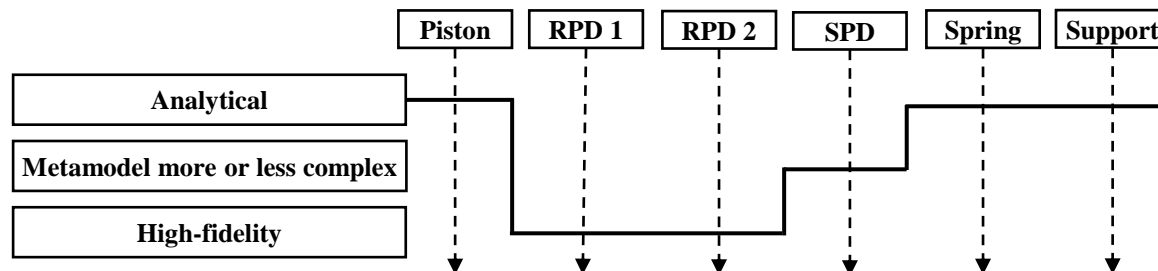
Thesis objectives

To develop Multi-fidelity modelling strategy based on a system approach including NURBS metamodels to solve engineer system model.

- To reduce computing time;
- To do quick preliminary design;
- To apply computing resources just where it is necessary.



- Idea: to **subdivide the system into elementary blocks** and to determine the **critical components** (pressure drop components).
- Friction coefficient (oil/piston): regular pressure drop → several level of physical description are available (Bernoulli / Metamodel / Finite Elements);
- Our goal: to evaluate different **robust** and **adaptative precision metamodeling** strategies.



The means (work in progress)

- System divide in **elementary blocks**:
 - **Modelica**
- To our typical problems, sensitivity analysis could be useful
 - **Sobol indices**
- Inclusion (FMI norm) of **external code or metamodel**:
 - NURBS, CATIA, Abaqus...
- Determination of the sufficient level of description:
 - **L^2 norm convergence** (still under reflexion)

What's new ?

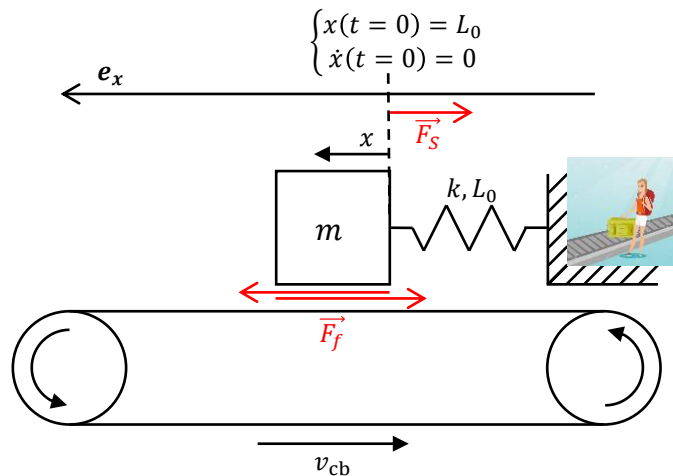
- To do multi-fidelity model with a system approach;
- To use Modelica coupled with other codes;
- To evaluate NURBS / PGD as a metamodel strategy.

Methodology illustration

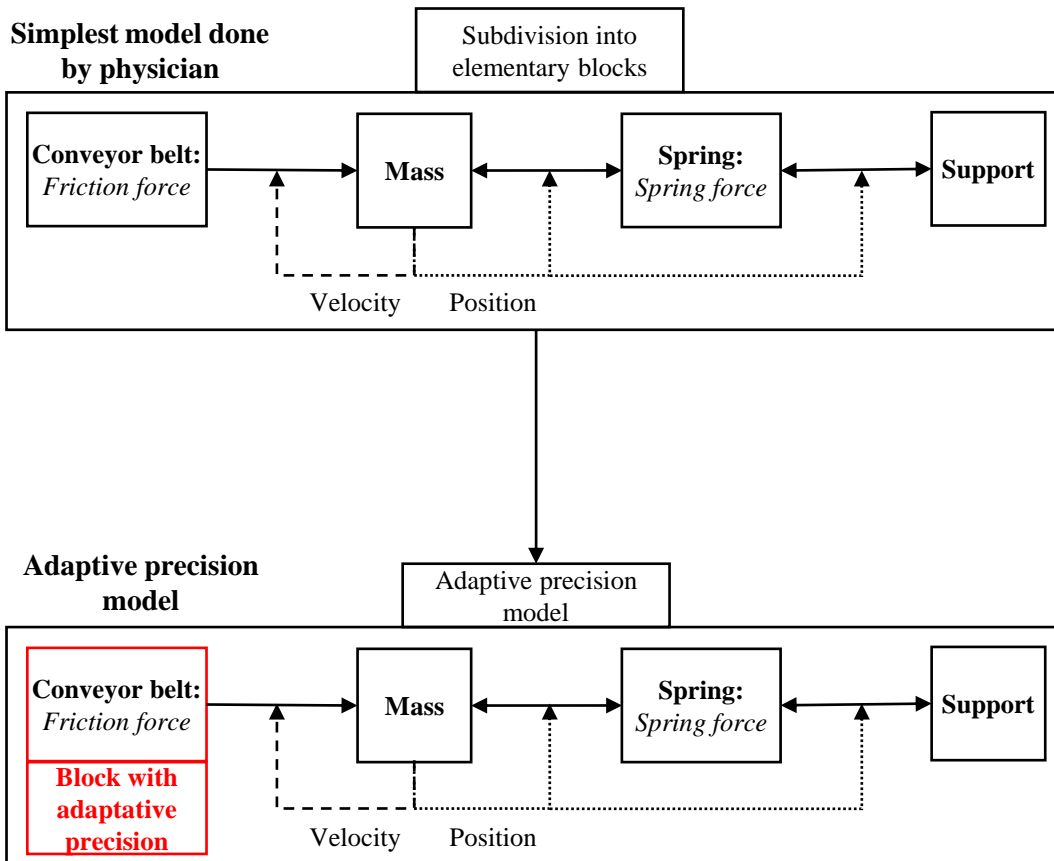
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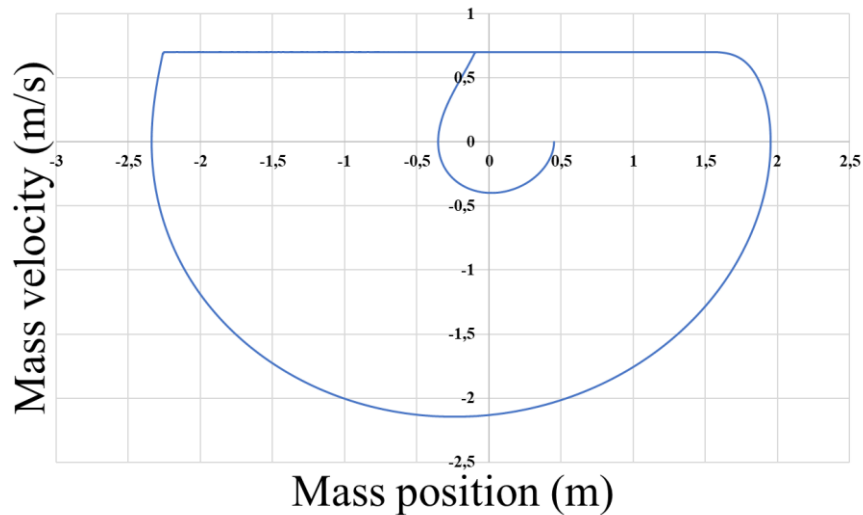
[2], [3], [4], [5]



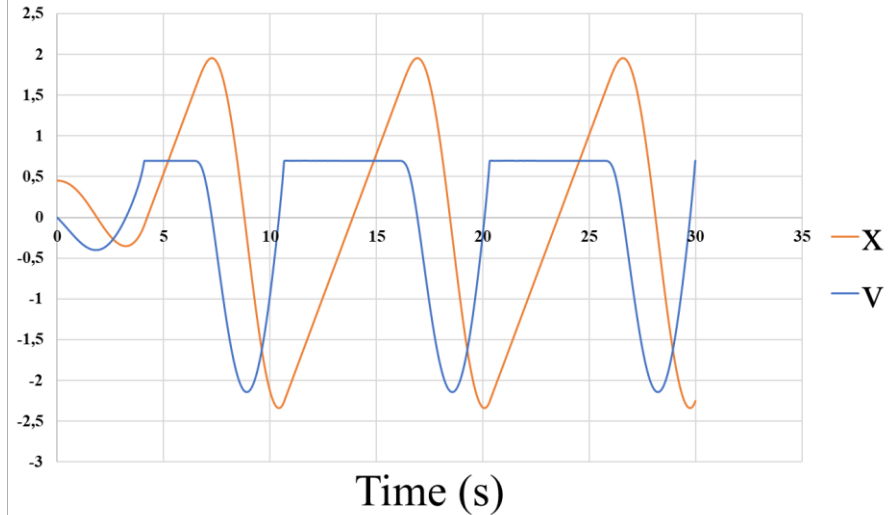
Scheme



Functional scheme



Phase diagram



Velocity and position vs time

To be sure to get the expected solution, a physician advice is required because the problem is highly non linear (cf. bibliography).

For sensibility analysis of the friction force block, we will consider the stable solution (velocity plateau > 0.3 s)

[14], [15], [16]

The Sobol's indices

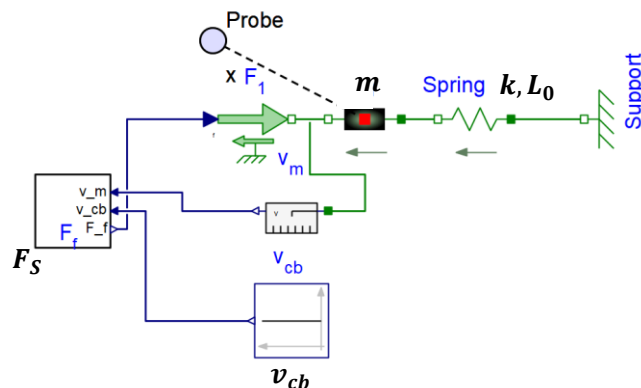
- Elementary indices translate the influence of one parameter on the observed output:

$$S_{E_i} = \frac{\text{Var}(\mathbb{E}(Y|X_i))}{\text{Var}(Y)}$$

- Total indices translate the influence of one parameter and its interactions with other parameters:

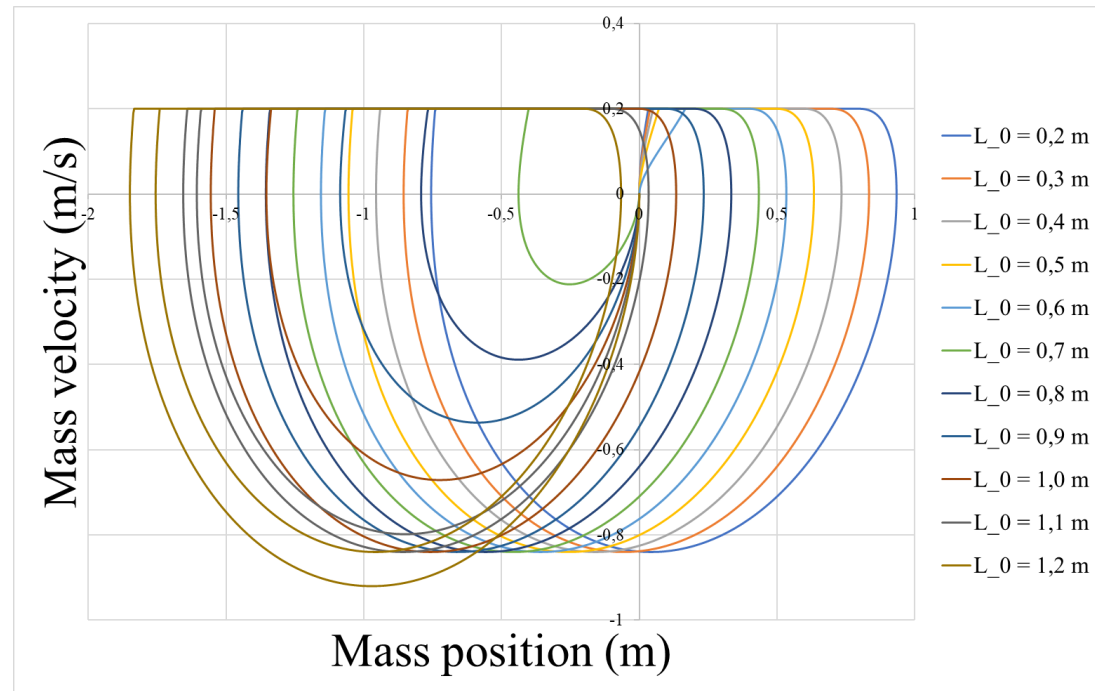
$$S_{T_i} = S_{E_i} + S_{(i,j)} + S_{(i,j,k)} + \dots$$

Example : motion stability (see previous solution)



Studied parameters	Range variation
k	$\mathcal{U}([0.5; 5.0]) \text{ N. m}^{-1}$
L_0	$\mathcal{U}([0.2; 1.2]) \text{ m}$
m	$\mathcal{U}([0.5; 2.0]) \text{ kg}$
v_{cb}	$\mathcal{U}([0.2; 5.0]) \text{ m. s}^{-1}$
F_s	$\mathcal{U}([1.0; 5.0]) \text{ N}$

Studied parameters	Elementary indices	Total indices
k	12.69 %	44.25
L_0	0.00 %	2.08
m	0.00 %	5.67
v_{cb}	34.05 %	74.69
F_S	11.47 %	44.19
Interactions	41.79 %	-



- Conveyor belt velocity is the most influential parameter.
- The unstretched spring length L_0 is the less influential parameter.
- We have to describe more precisely the elementary block which contains v_{cb} and F_S (the frictional force block).

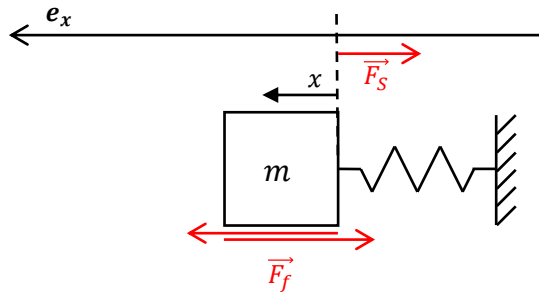
The means

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[11], [12], [13]

The computer language: system approach and acausal resolution



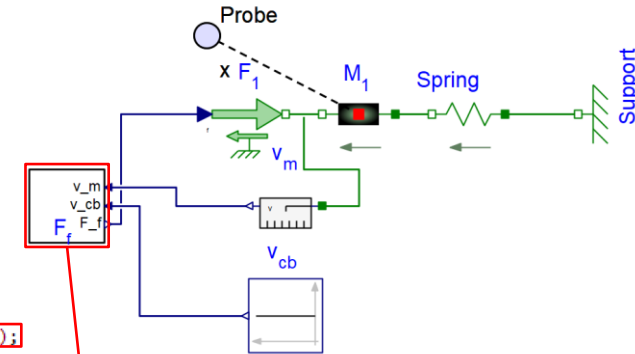
Real model

```

1 model StickSlip
2   import SI = Modelica.SIunits;
3   parameter SI.SpringConstant k = 1;
4   parameter SI.Length L0 = 0.45;
5   parameter SI.Force F_S = 2;
6   parameter SI.Mass m=1;
7   parameter SI.Velocity v_cb = 0.2;
8   parameter Real delta = 3;
9   SI.Position x;
10  SI.Velocity v, v_rel;
11  SI.Acceleration a;
12  SI.Force F_Spring, F_f;
13  equation
14  v = der(x);
15  a = der(v);
16  F_Spring = k*(x-L0);
17  F_f = -(F_S*sign(v_rel))/(1 + delta * abs(v_rel));
18  v_rel = v - v_cb;
19  m*a = F_Spring + F_f;
20 end StickSlip;

```

Computed model



Scheme block

Metamodeling
strategy

Modelica vs. classical language:

In Modelica: write just the constraint equations

Classical language: write in an explicit way the final Differential Algebraic Equation system

Identification of some physical problems of interest: thermo-mechanical system, latches system, thermal battery

Choice of metamodeling strategy

- The importance of the *off-line* part;
- The difficulty to use it;
- The difficulty to integrate it;
- Their precision;
- How it take in account the uncertainty.

What I will evaluate in my thesis

- NURBS
 - Approximation strategy, improved by the I2M team (genetic algorithm), especially during the thesis of Yohann AUDOUX.
- PGD
 - Model reduction method (contact with Francisco CHINESTA [10]).

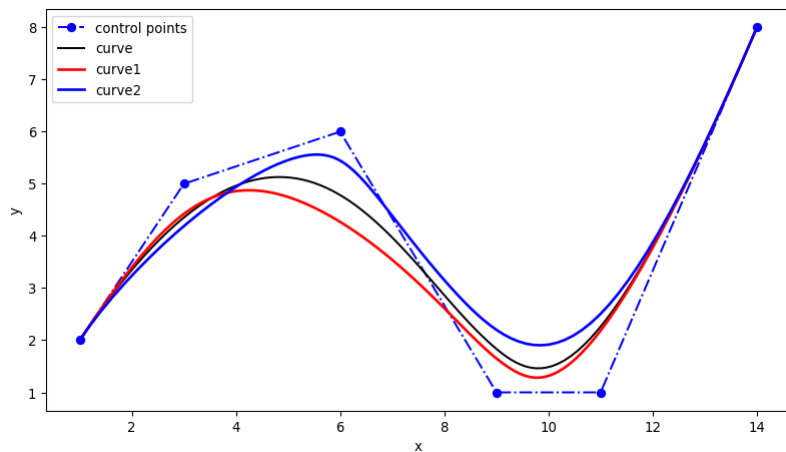
[6], [8]

NURBS curves (Non Uniform Rational Basis Spline)

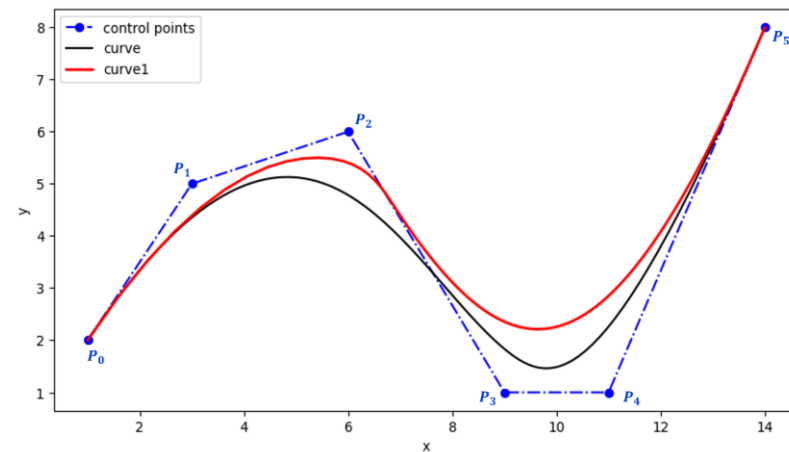
$$\mathbf{C}(t) = \frac{\sum_{i=0}^{p-1} N_i^n(t) \omega_i \mathbf{P}_i}{\sum_{i=0}^{p-1} N_i^n(t) \omega_i}$$

- p : the number of control points \mathbf{P}_i (forming a control polygon);
- ω_i : the weights associated to each control points \mathbf{P}_i ;
- t : node of the node vector $\mathbf{t} = [t_0, \dots, t_m]$, with $m = n + p + 1$;
- $N(t)$: parametrical functions of degree n , defined on $t \in [t_k; t_{k+1}]$ with $k \in [0; m - 1]$ and expressed recursively as follows:

$$\begin{cases} N_{i,0}(t) = \begin{cases} 1, & \text{if } t_i < t < t_{i+1}, \\ 0, & \text{else,} \end{cases} \\ N_{i,\tau}(t) = \frac{t - t_i}{t_{i+\tau} - t_i} N_{i,\tau-1}(t) + \frac{t_{i+\tau+1} - t}{t_{i+\tau+1} - t_{i+1}}, \end{cases} \quad \text{with } i \in [0; p - 1] \text{ and } \tau \in [0; n].$$



In black: simple B-Spline;
 In red: high weight on node 3;
 In blue: low weight on node 3.



In black: simple B-Spline with uniform node vector
 $\mathbf{t} = [0; 0; 0; 0; \frac{1}{3}; \frac{2}{3}; 1; 1; 1; 1];$
 In red: NURBS with non-uniform node vector
 $\mathbf{t} = [0; 0; 0; 0; 0.1; 0.2; 1; 1; 1; 1].$

[7], [9]

Use in metamodeling strategy:

- Pioneer in this field: Turner
- Algorithm developed and generalised at I2M
- Formalised by Yohann Audoux during his thesis
 - Use of a NURBS algorithm coupled with a genetic optimisation algorithm to compute the node vector and point weight values.
- **Can be included in Modelica as all functions with several parameters**

Conclusion and perspectives

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What has been done (*1st PhD year*)

- Identification of some systems application;
- First sensitivity analysis;
- Physical / engineering analysis (by the user) + Modelica;
- Bibliography: Modelica and its uses/viability, metamodeling, NURBS, PGD.

Future work (*2nd PhD year*)

- To evaluate NURBS and PGD programs;
- To compare others metamodel strategies with NURBS and PGD;
- To take into account: uncertainty, multi-fidelity (to determine the good level of description) [1]



Thank for your attention

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Multi-fidelity:

[1] Loïc Le Gratier, Claire Cannamela. Kriging-based sequential design strategies using fast cross-validation techniques with extensions to multi-fidelity computer codes. 2012.

Stick-Slip model:

[2] Sieber, J., & Bernd, K. (2008). Control based bifurcation analysis for experiments. 18.

[3] Leine, R. I., Van Campen, D. H., De Kraker, A., & Van Den Steen, L. (1998). Stick-Slip Vibrations Induced by Alternate Friction Models. *Nonlinear Dynamics*. <https://doi.org/10.1023/A:1008289604683>

[4] Dieci, L., & Lopez, L. (2009). Sliding motion in filippov differential systems: Theoretical results and a computational approach. *SIAM Journal on Numerical Analysis*. <https://doi.org/10.1137/080724599>

[5] Dieci, L., & Lopez, L. (2012). A survey of numerical methods for IVPs of ODEs with discontinuous right-hand side. *Journal of Computational and Applied Mathematics*. <https://doi.org/10.1016/j.cam.2012.02.011>

NURBS:

[6] Audoux, Y. (2019). Développement d'une nouvelle méthode de réduction de modèle basée sur les hypersurfaces NURBS (Non-Uniform Rational Basis Splines). Arts et Métiers ParisTech - Institut de Mécanique et de l'Ingénierie.

[7] De Boor, C. (1972). On calculating with B-Spline. *Journal of Approximation Theory*, 6(1), 50–62.

[8] Turner, C. J. (2012), Robust optimization of mixed-integer problems using NURBS-Based Métamodels, *J. Comput. Inf. Sci. Eng.* Dec 2012, 12(4): 041010 (7 pages)

[9] Piegl Les and Tiller Wayne. *The NURBS Book*. Springer-Verlag, New York, NY, USA, second edition, 1996.

PGD:

[10] Ammar, A., Mokdad, B., Chinesta, F., & Keunings, R. (2007). A new family of solvers for some classes of multidimensional partial differential equations encountered in kinetic theory modelling of complex fluids. *Journal of Non-Newtonian Fluid Mechanics*, 144(2–3), 98–121.

Modelica:

[11] Peter Fritzson. *Principles of Object Oriented Modeling and Simulation with Modelica 3.3 : A Cyber-Physical Approach*. 2014.

[12] Sven Erik Mattsson and Gustaf Söderlind. Index Reduction in Differential-Algebraic Equations Using Dummy Derivatives. *SIAM Journal on Scientific Computing*, 14(3) :677–692, 1993.

[13] Torsten Blockwitz, Martin Otter, Johan Akesson, Martin Arnold, Christoph Clauss, Hilding Elmqvist, Markus Friedrich, Andreas Junghanns, Jakob Mauss, Dietmar Neumerkel, Hans Olsson, and Antoine Viel. *Functional Mockup Interface 2.0 : The Standard for Tool independent Exchange of Simulation Models*. 2012.

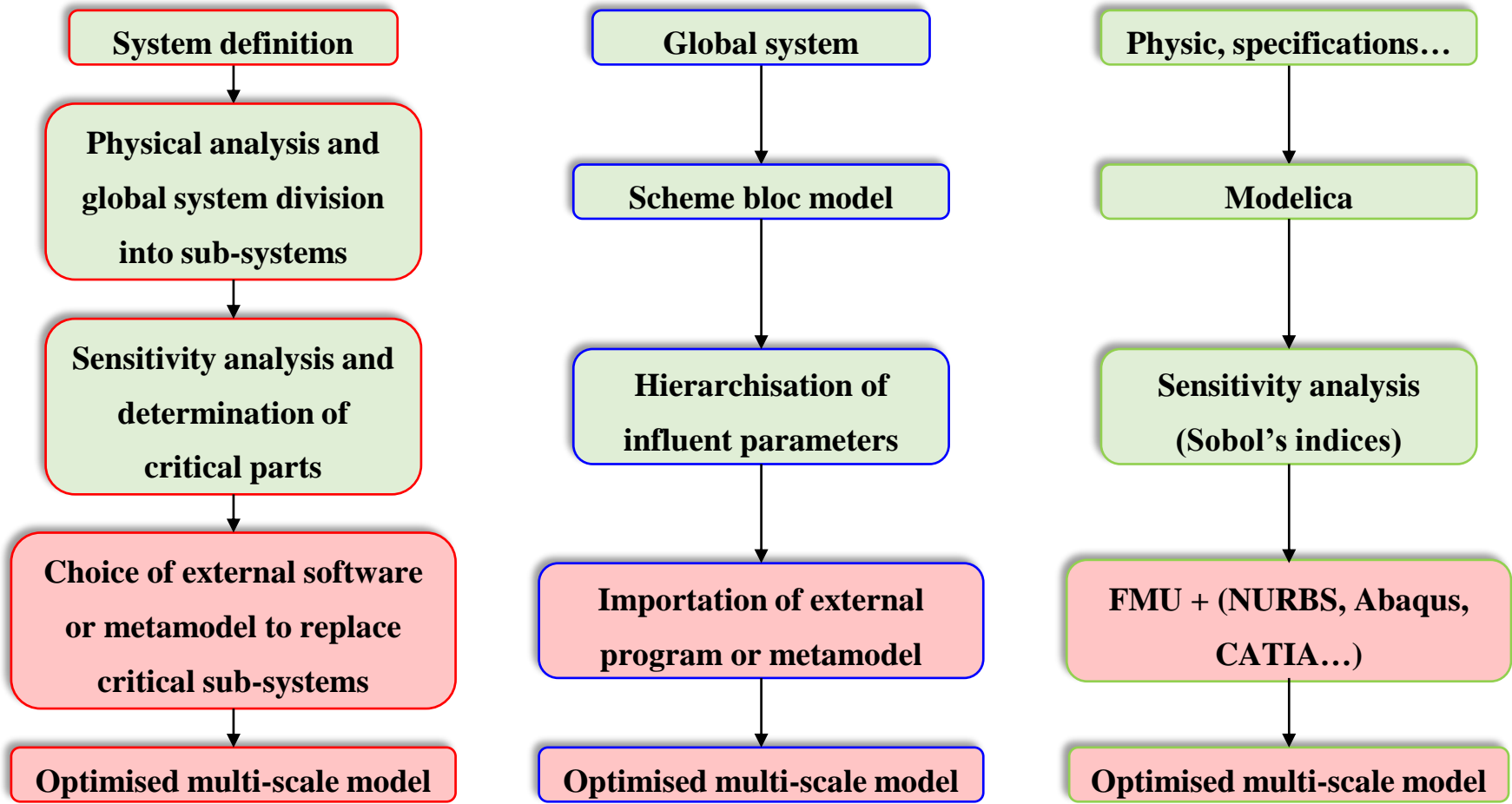
Sobol:

[14] Ilya Meïérovitch Sobol. Sensitivity Estimates for Nonlinear Mathematical Models. *MMCE*, 1(4) :407–414, 1993.

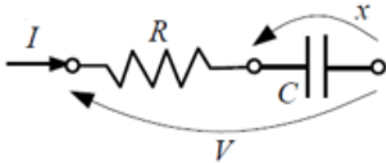
[15] Maxime Heliot. Studies and development of a sensitivity analysis tool and application to an aerothermal model. Master's thesis, Cranfield University, 2018.

[16] Iooss, B., & Lemaître, P. (2015). A review on global sensitivity analysis methods. *Operations Research/ Computer Science Interfaces Series*, 59, 101–122. https://doi.org/10.1007/978-1-4899-7547-8_5

Summary of the methodology developed through my thesis



The computer language

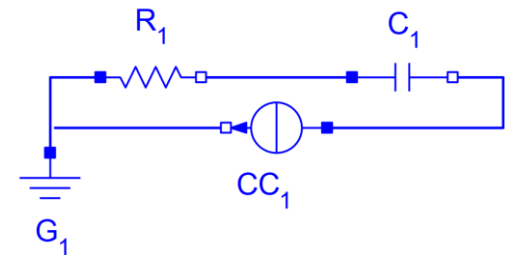


Real model

```

1 model CircuitRC
2   import SI = Modelica.SIunits;
3   parameter SI.Current I=1;
4   parameter SI.Resistance R=1;
5   parameter SI.Capacitance C=1;
6   SI.Voltage V, x(start = 0);
7   equation
8     x + R*I = V;
9     C * der(x) = I;
10  end CircuitRC;
  
```

Computed model



Bloc scheme

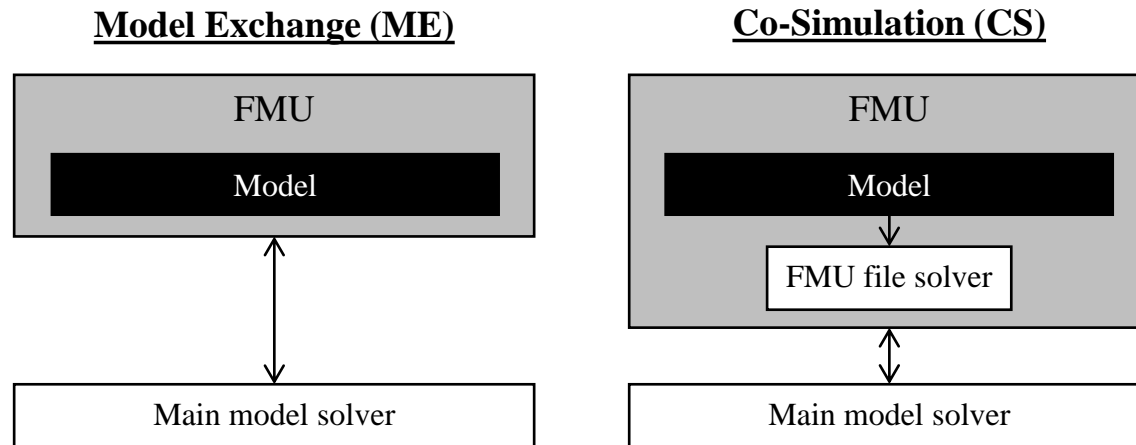
$$\begin{cases} x + RI = V \\ C\dot{x} = I \end{cases}$$

Initial system

$$C = \frac{I}{\dot{x}} \text{ ou } R = \frac{V-x}{I}$$

Equations which could
be taken into account

The Functional Mock-up Interface norm (FMI)



A FMU file is like a ZIP file which contains:

- An XML document with the description of the model
- A C/C++ code of the model
- Other documents that the model needs to be run (Excel table point for example)

The elementary Sobol's indices

- It corresponds to the influence of a parameter translated by **its variance, normalized** on the **global model variance**.
- Explanation :
 - The parameter ξ_i variance is determined by studying the output variance Y by fixing $\xi_i : Var(Y|\xi_i = x_i) \xrightarrow{\text{global method}} E(Var(Y|\xi_i))$. This corresponds to a **conditional variance** or **conditional expectation**.
 - If $E(Var(Y|\xi_i))$ is small, fixing ξ_i reduces the variability of Y , so ξ_i is influential.
 - By using the total variance theorem : $Var(Y) = E(Var(Y|\xi_i)) + Var(E(Y|\xi_i))$, it can be deduced that if ξ_i is influential, $Var(E(Y|\xi_i))$ is high.
 - Normalizing by $Var(Y)$, we obtain the influence of the parameter ξ_i on the output Y in percent, which is much more exploitable. This is **the elementary Sobol index** S_{E_i} .

$$S_{E_i} = \frac{Var(E(Y|\xi_i))}{Var(Y)} = \frac{E\left(\left(E(Y|\xi_i) - E(Y)\right)^2\right)}{Var(Y)}$$

The total Sobol's indices

- It corresponds to the influence of a parameter alone and with its interactions with other parameters translated by **its variance** and **coupled variances**, **normalized** on the **global model variance**.
- Explanation:
 - A parameter can have a self influence on the system and cross-influence with another parameter.
 - To take it in case, we use the **total Sobol's index** which correspond to the sum of elementary Sobol's index, second order Sobol's index, third order...
 - A Sobol's index of order greater than one, is determined by fixing the value of more than one parameter. Mathematically, this is expressed has follows:

$$S_{E_{(i,j)}} = \frac{1}{\text{Var}(Y)} \left[\text{Var} \left(E(Y|\xi_i, \xi_j) \right) - \text{Var}(E(Y|\xi_i)) - \text{Var} \left(E(Y|\xi_j) \right) \right]$$

$$S_{E_{(i,j,k)}} = \frac{1}{\text{Var}(Y)} \left[\text{Var} \left(E(Y|\xi_i, \xi_j, \xi_k) \right) - \text{Var}(E(Y|\xi_i)) - \text{Var} \left(E(Y|\xi_j) \right) - \text{Var}(E(Y|\xi_k)) \right]$$

- Finally, by summing all the Sobol's index of order one and more, we deduce the **total Sobol's index** :

$$S_{T_i} = S_{E_i} + S_{E_{(i,j)}} + S_{E_{(i,j,k)}} + \dots$$

Sobol's index property

- By considering a model of d parameters, we have the following low:

$$\sum_{i=1}^d S_{E_i} + \sum_{\substack{i=1, j=1 \\ i \neq j}}^d S_{E_{(i,j)}} + \sum_{\substack{i=1, j=1, k=1 \\ i \neq j \neq k}}^d S_{E_{(i,j,k)}} + \dots = 1$$

Caution when using this indicator

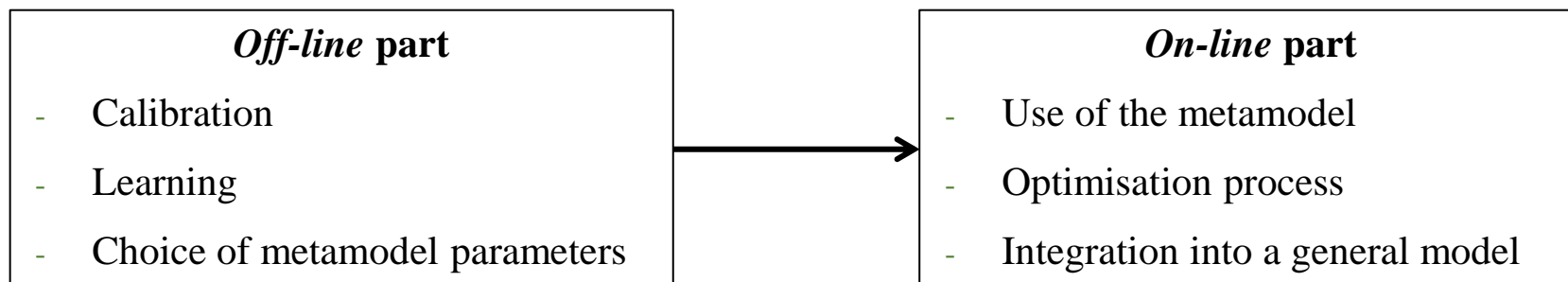
It's important to well make the difference with elementary and total Sobol's index. Indeed, a parameter can not have influence alone but can have great with interactions. So, it's necessary to first study total Sobol's index and after elementary.

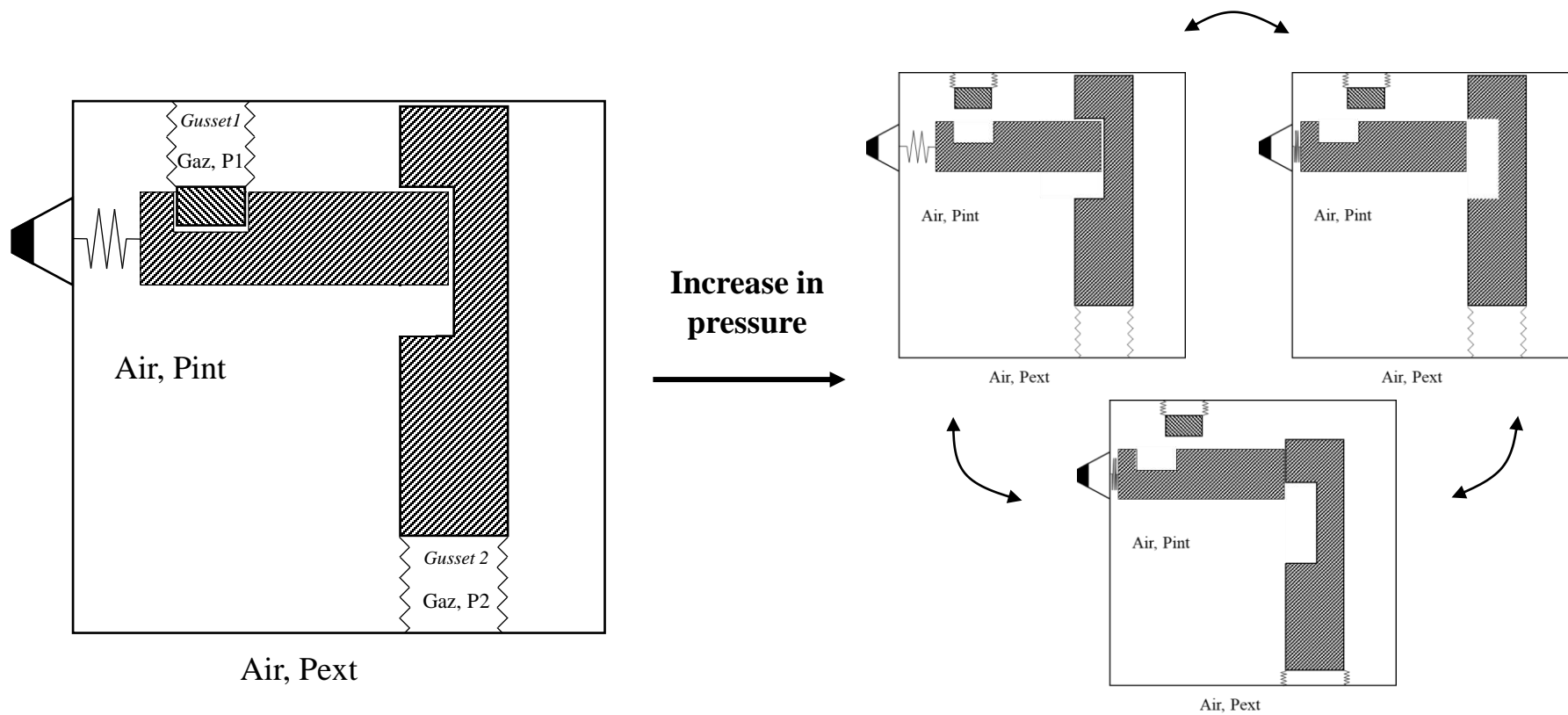
Model definition

Input parameters X	Studied system	Output variables $f(X)$
- Physical characteristics	- Mathematical system	- Movements
- Chemical characteristics	- Digital simulation	- Stability
- ...	- ...	- ...

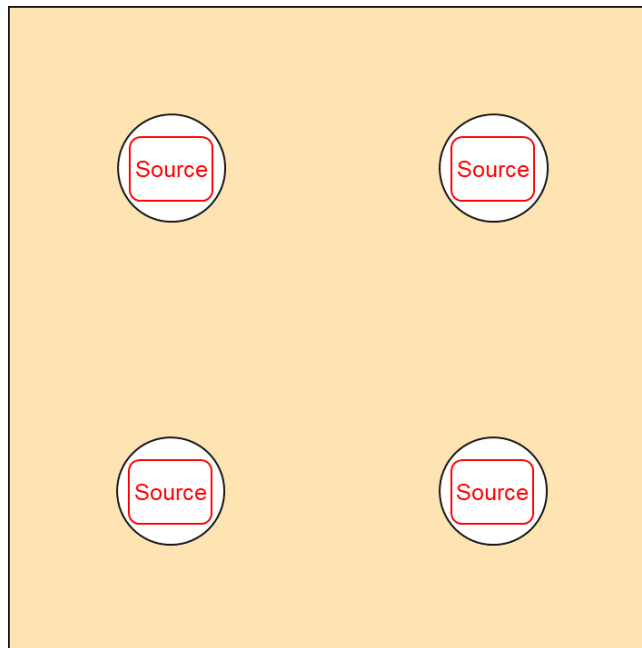
A metamodeling strategy will be to determine a function $\hat{f}(X)$ which will approximate the initial function $f(X)$ using fewer resources and with the lowest possible approximation error ϵ .

There are 2 parts in metamodeling

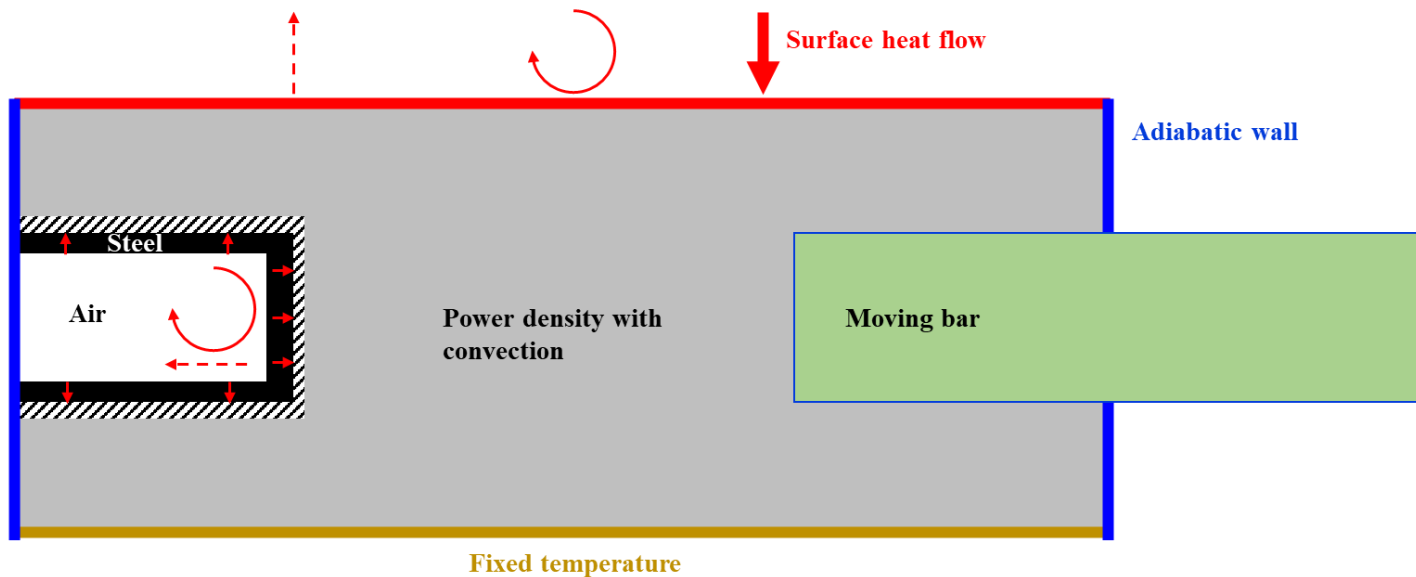




Criteria	Comments
Physic(s)	Hydromechanical
Metamodel – microscopical	Pressure law
Analytical model – macroscopical	Latches motions



Criteria	Comments
Physic(s)	Thermal
Metamodel – microscopical	Phase change material
Analytical model – macroscopical	Heat exchanges



Criteria	Comments
Physic(s)	Thermomechanical
Metamodel – microscopical	Material with volume power (thermal) Air cavity (thermal)
Analytical model – macroscopical	Heat transfer Movement of the bar