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The Reduced Basis Method Applied to Aerothermal Simulations

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The Reduced Basis Method

Motivations

- Modeling : multi-physics non-linear models, complex geometries, genericity
- Uncertainty management / Risk analysis
- Optimization in early design, certification or operating phases

Objective 1: Fast

- Complex geometries
- \longrightarrow Large number of dofs
- Uncertainty quantification
- \rightarrow Large number of runs

• Field quality • Design optimization

 \longrightarrow Certified bounds

Objective 2: Reliable

 \longrightarrow Reach material limits

Inlets Outlets Sensor locations Passengers 0 0 00 Equipments

Airbus Use-Case

Propose in the context of the ANR Project CHORUS **Objective** : Apply the Certified Reduced Basis Methods on an aerothermal simulation in an avionic bay

Model :

- Steady Navier-Stokes/Heat transfer
- Incompressible Newtonian Fluid
- Boussinesq Approximation
- Turbulent Flow

Main Idea

Weak formulation of the model : $a(u(\mu), v; \mu) = f(v; \mu)$

FEM Approximation:

$$\underbrace{\mathcal{K}^{\mathcal{N}} = span\{\phi_{1}, \dots, \phi_{\mathcal{N}}\}}_{\text{FEM approximation space}} \longrightarrow \underbrace{u^{\mathcal{N}}(\mu) = \sum_{i=1}^{\mathcal{N}} u^{\mathcal{N}}_{i} \phi_{i}}_{\mathcal{N} \approx 10^{6}} \longrightarrow \underbrace{\mathcal{A}^{\mathcal{N}}(\mu)u^{\mathcal{N}}(\mu) = F^{\mathcal{N}}(\mu)}_{\mathcal{N} \times \mathcal{N} \text{ system expensive to solve}}$$

RB Approximation: $u^N(\mu) \approx u^{\mathcal{N}}(\mu)$: linear combination of FEM solution

$$\underbrace{W^{N} = span\{u^{\mathcal{N}}(\mu_{1}), \dots, u^{\mathcal{N}}(\mu_{N})\}}_{\text{RB approximation space}} \rightarrow \underbrace{u^{\mathcal{N}}(\mu_{N}) = \sum_{i=1}^{N} u_{i}^{\mathcal{N}}(\mu) \underline{u}^{\mathcal{N}}(\mu_{i})}_{10 \leq N \leq 100} \rightarrow \underbrace{A^{\mathcal{N}}(\mu)u^{\mathcal{N}}(\mu) = F^{\mathcal{N}}(\mu)}_{\mathcal{N} \times N \text{ system cheaper to solve}}$$

Ingredients

- Set of parameters : $\mathcal{D}^{\boldsymbol{\mu}}$
- FEM 'truth" approximation $X^{\mathcal{N}}$: finite element approximation space of dimension $\mathcal{N} >> 1$ $u^{\mathcal{N}}(\mu) \in X^{\mathcal{N}}$ is solution of $a(u^{\mathcal{N}}(\mu), v; \mu) = f(v; \mu) \quad \forall v \in X^{\mathcal{N}}$
- RB approximation

Sample : $S_N = \{ \mu_1 \in \mathcal{D}^{\boldsymbol{\mu}}, ..., \mu_N \in \mathcal{D}^{\boldsymbol{\mu}} \}$ Approximation space : $W_N = span\{u^N(\mu_1), ..., u^N(\mu_N)\}$ with N << NGalerkin projection on W_N to determine RB coefficients

Efficient offline-online strategy

$$u^{\mathcal{N}}(oldsymbol{\mu}) = \sum_{i=1}^{\mathcal{N}} u^{\mathcal{N}}_i(oldsymbol{\mu}) \overline{u}^{\mathcal{N}}(oldsymbol{\mu}_i)$$

 $N \times N$ system to solve : $\sum_{i=1}^{N} a(\underline{u}^{\mathcal{N}}(\mu_i), v_k; \mu) u_i^{\mathcal{N}}(\mu) = f(v_k; \mu), \ 1 \leq k \leq N$ If the parameter dependance can be

Mathematical Model

Governing Equations

$$\begin{cases} \rho \mathbf{u} \cdot \nabla \mathbf{u} + \nabla p - \mu \Delta \mathbf{u} = \rho \beta (T - T_0) g, & \text{in } \Omega \times [0, T_f], \\ \nabla \cdot \mathbf{u} = 0, & \text{in } \Omega \times [0, T_f], \\ \mathbf{u} \cdot \nabla T - \kappa \Delta T = 0, & \text{in } \Omega \times [0, T_f], \\ + \text{Boundary Conditions.} \end{cases}$$

with

• ρ : fluid density $(kg.m^{-3})$, • $\kappa \approx 10^{-5} m^2.s^{-1}$: thermal diffusivity $(m^2.s^{-1})$, • $\mu \approx 10^{-5} m^2.s^{-1}$: dynamic viscosity $(kg.m^{-1}.s^{-1})$,

• β : thermal expansion coefficient (K^{-1}), • Length Scale $L \approx 1~m$, • $\Rightarrow Re \approx 10^5$, $Pr \approx 1$, $Pe \approx 10^5$.

(1)

Solving Strategy

- Finite Element Discretization
- Newton Algorithm with transient continuation
- Parallel implementation using Feel++ library: http://www.feelpp.org/

Challenge and Difficulties

- Multi-physic coupled model: simultaneous construction of the different reduced spaces
- High Reynolds flow: use of stabilization methods (SUPG/GLS) in the FEM and the RB model
- Non-Linearity: Newton algorithm with an affine decomposition of the Jacobian/Residual
- Non-affine terms: use of Empirical Interpolation Method (EIM) for discrete operators
- Non-Linearity: Use of Simultaneous EIM and RB (SER) algorithm to generate an affine approximation of the

expressed as an affine decomposition :



- non-linear terms (stabilization terms)
- Complex Formulation: Due to geometric parameters, use of EIM to automatically recover the affine decomposition.

Perspectives

- Development of Efficient Error Estimators for the Reduced Model
- Reduction of the Coupled Turbulence Model

Numerical Results: Cooling of an Printed Circuit Board, Reduced Model



2D model representative of the neighboring of an electronic component submitted to a cooling air flow.

• Air thermal diffusivity: $\kappa_a = 2.7 \cdot 10^{-5}$ • Air kinematic viscosity: $\mu_a = 1.9 \cdot 10^{-5}$

Parameters

- Q_1 and Q_2 : Heat sources from the two Integrated Circuits: [0, 10⁶]
- κ_1 and κ_2 : Thermal conductivity of the two Integrated Circuits: [0.2, 150]
- D: The inflow rate: $[5 \cdot 10^{-4}, 10^{-2}]$



Reduced-basis method (rbm) for non-affine elliptic parametrized pdes.(phd).

Ulm University, 2012.





Sponsor	Références
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The main objective of this project is to construct an e-infrastructure that provides, in a user-driven, integrative way, tailored access to the necessary	Efficient model reduction of parametrized systems by matrix discrete empirical interpolation. Journal of Computational Physics, 303:431–454, 2015. T Tonn.

services, resources and even tools for the fast prototyping, providing the service producers with the mathematical frameworks as well.

