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The Reduced Basis Method Applied to Aerothermal Simulations

JB. Wahl, C. Prud'homme, Y. Hoarau



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
→ Large number of dofs
- Uncertainty quantification
→ Large number of runs

Objective 2: Reliable

- Field quality
- Design optimization
→ Certified bounds
- Reach material limits

Main Idea

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

FEM Approximation:

$$X^N = \text{span}\{\phi_1, \dots, \phi_N\} \xrightarrow[\text{FEM approximation space}]{} u^N(\mu) = \sum_{i=1}^N u_i^N \phi_i \xrightarrow[N \approx 10^6]{} A^N(\mu) u^N(\mu) = F^N(\mu)$$

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

$$W^N = \text{span}\{u^N(\mu_1), \dots, u^N(\mu_N)\} \xrightarrow[\text{RB approximation space}]{} u^N(\mu) = \sum_{i=1}^N u_i^N(\mu) \underline{u}^N(\mu_i) \xrightarrow[10 \leq N \leq 100]{} A^N(\mu) u^N(\mu) = F^N(\mu)$$

Ingredients

- Set of parameters : \mathcal{D}^μ
- FEM 'truth' approximation
 X^N : finite element approximation space of dimension $N >> 1$
 $u^N(\mu) \in X^N$ is solution of $a(u^N(\mu), v; \mu) = f(v; \mu) \quad \forall v \in X^N$
- RB approximation
Sample : $S_N = \{\mu_1 \in \mathcal{D}^\mu, \dots, \mu_N \in \mathcal{D}^\mu\}$
Approximation space : $W_N = \text{span}\{u^N(\mu_1), \dots, u^N(\mu_N)\}$ with $N \ll N$
Galerkin projection on W_N to determine RB coefficients

Efficient offline-online strategy

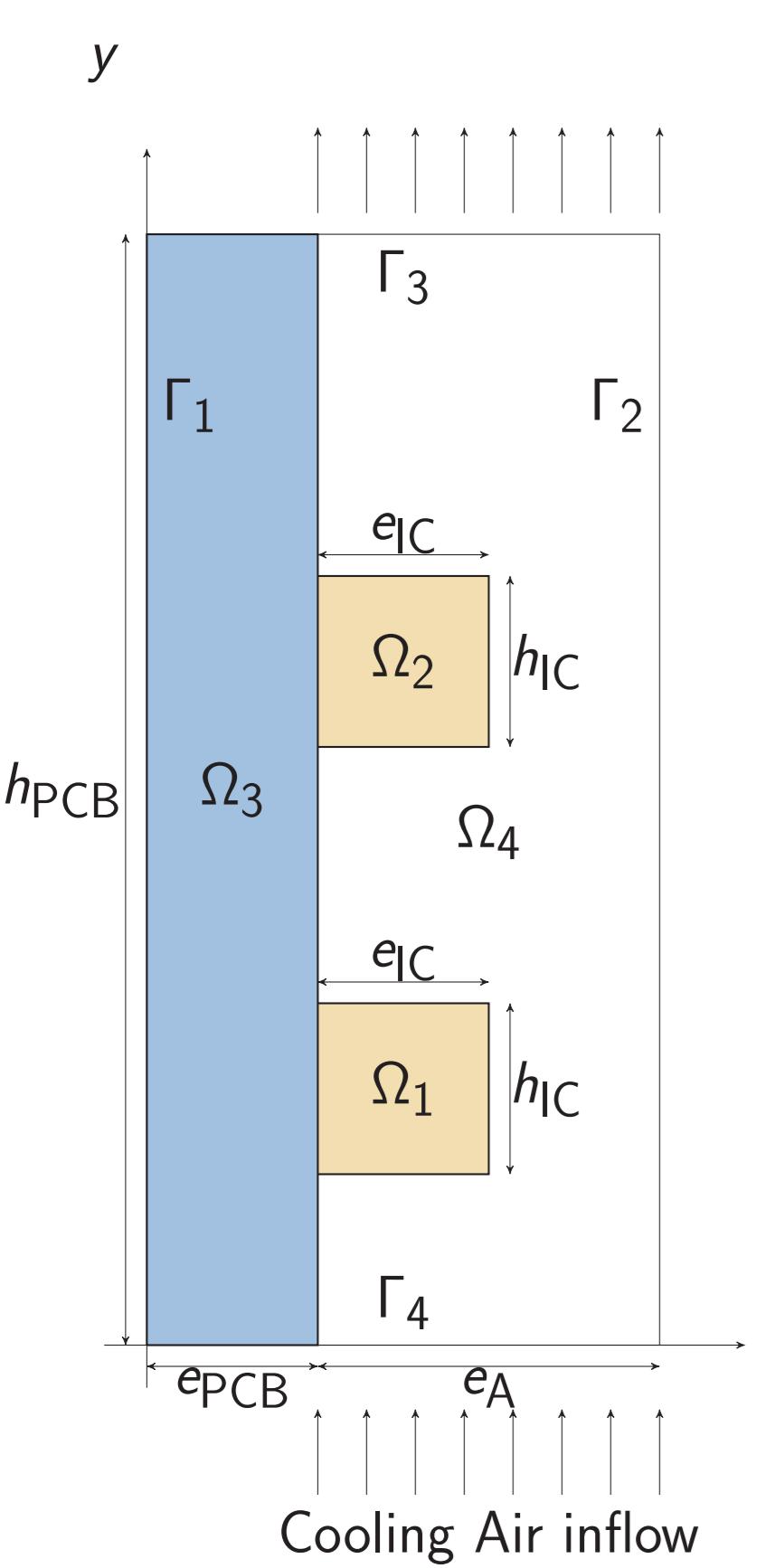
$$u^N(\mu) = \sum_{i=1}^N u_i^N(\mu) \underline{u}^N(\mu_i)$$

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

expressed as an affine decomposition :

$$\begin{aligned} a(u, v; \mu) &= \sum_q^{Q_a} \theta_q^a(\mu) a_q(u, v) \quad \text{and} \quad f(v; \mu) = \sum_q^{Q_f} \theta_q^f(\mu) f_q(v) \\ \Rightarrow \sum_{i=1}^N \left[\sum_{q=1}^{Q_a} \theta_q^a(\mu) \underbrace{a^q(\underline{u}^N(\mu_i), \underline{u}^N(\mu_j))}_{\text{precomputed}} \right] u_i^N(\mu) &= \sum_{q=1}^{Q_f} \theta_q^f(\mu) \underbrace{f^q(\underline{u}^N(\mu_j))}_{\text{precomputed}} \end{aligned}$$

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

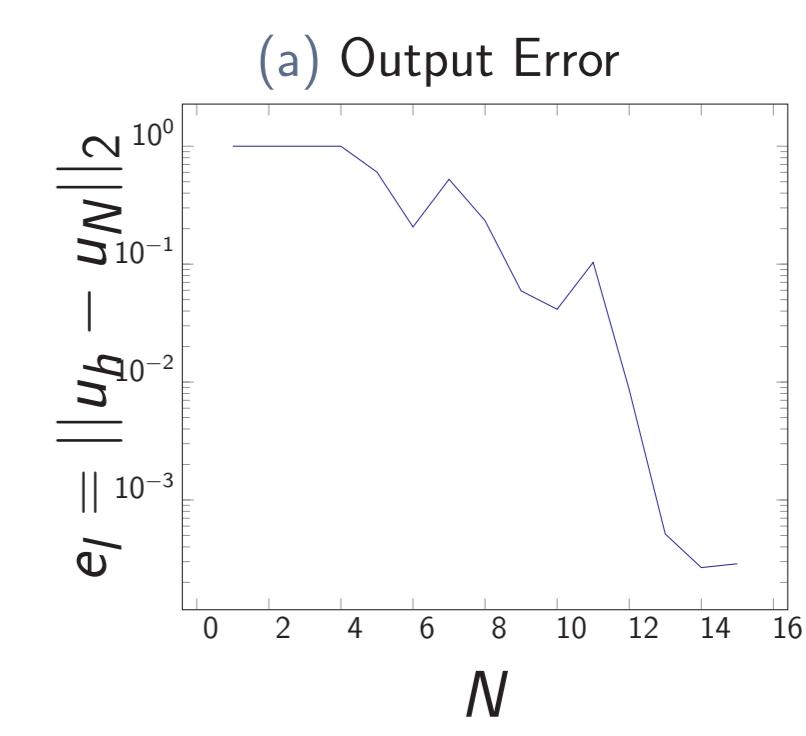
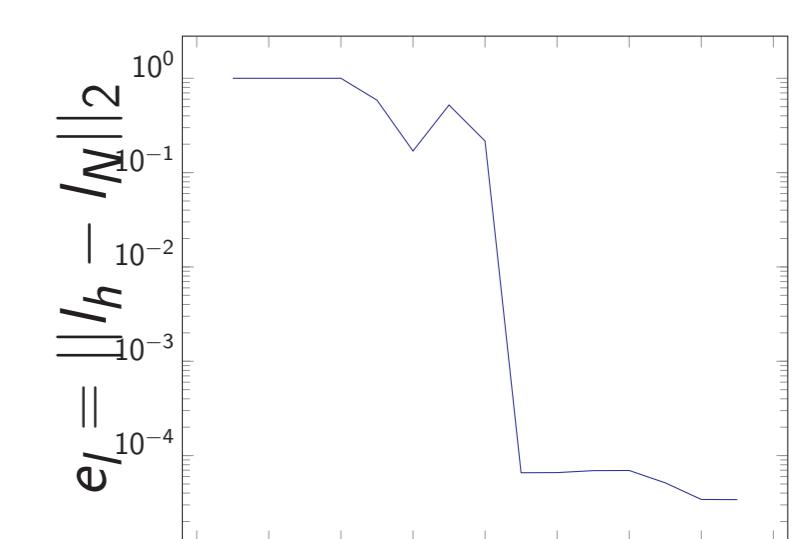
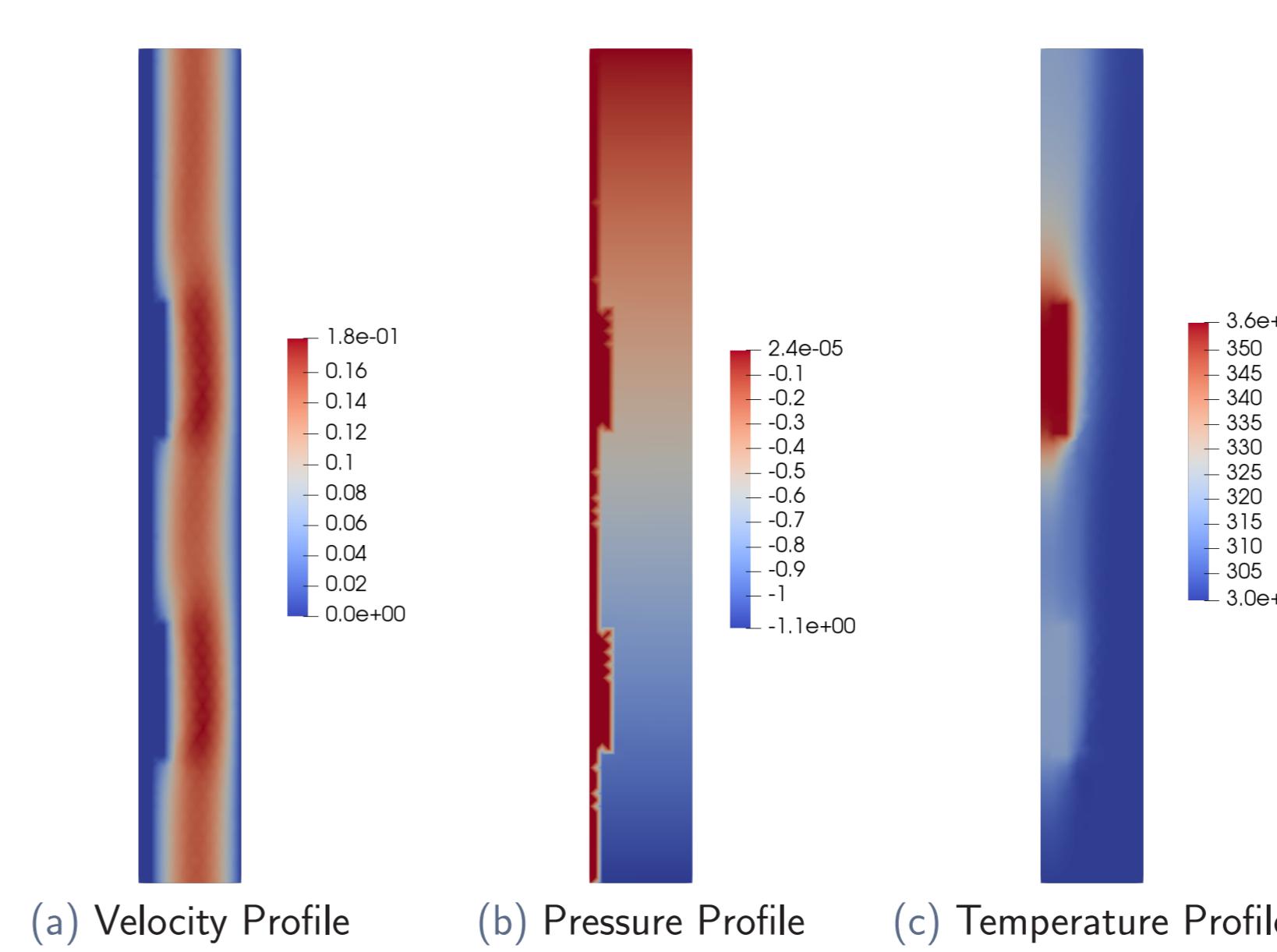


Physical Model

- 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: $[0, 10^6]$
- κ_1 and κ_2 : Thermal conductivity of the two Integrated Circuits: $[0.2, 150]$
- D : The inflow rate: $[5 \cdot 10^{-4}, 10^{-2}]$



Convergence with respect to the size of the basis.
Maximum of the error, evaluated on 100 RB approximations compared with the FEM approximations

Sponsor



MSO4SC:
Mathematical Modeling, Simulation and Optimization
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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 services, resources and even tools for the fast prototyping, providing the service producers with the mathematical frameworks as well.

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