

Non Intrusive Reduced Basis method (NIRB) The Two-grids method

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Introduction

The two-grids method is non intrusive



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NIRB method

Offline

Online

Results with FE solver

Finite volume solver

Error estimate

Results with FV solver

Applications

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Conclusions and perspectives



Industrial context → **black box solver (BB)**

Non intrusive reduced basis method useful for:

- Optimization parameters fitting
- High fidelity real-time simulations

Goal: Solve for several parameters the same parameter dependent problem and reduce the computational costs

Several methods:

- **Finite Element method**
- **Extension to Finite Volume method**

Two examples of applications



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■ Offshore wind Farm

A way to reduce computational costs of an offshore wind farm to optimize the position of the wind turbines.

■ Spatial Variability

What if we could consider a non physically acceptable truncated domain?

A model problem



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$$\begin{cases} -\operatorname{div}(\mathbf{A}(\boldsymbol{\mu})\nabla u) = f & \text{in } \Omega, \\ u = 0 & \text{on } \partial\Omega, \end{cases} \quad \begin{matrix} (1a) \\ (1b) \end{matrix}$$

- $u(\mathbf{x}; \boldsymbol{\mu})$: Unknowns (u_h on the fine mesh \mathcal{T}_h , u_H on the coarse mesh \mathcal{T}_H).
- $\boldsymbol{\mu} \in \mathcal{S}$: Variable parameter,
 $f \in L^2(\Omega)$,
 $\mathbf{A} : \mathcal{S} \times \Omega \rightarrow \mathcal{M}_d(\mathbb{R})$ is measurable, bounded, uniformly elliptic,
and $\mathbf{A}(\mathbf{x})$ is symmetric for a.e. $\mathbf{x} \in \Omega$.

NIRB scheme OFFLINE/ONLINE



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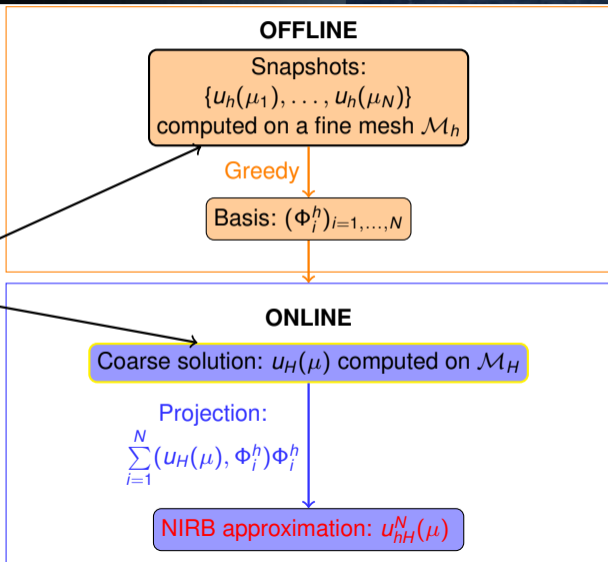
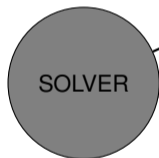
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⚠ Parameters
 $\{\mu_1, \dots, \mu_N\}$
must be
well chosen!



How to choose the parameters



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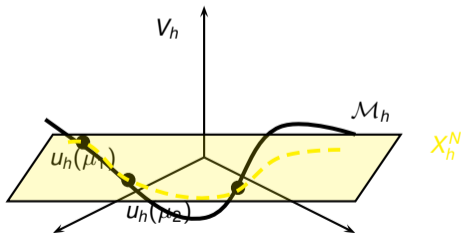
■ Greedy algorithm

■ Observing the decay of eigenvalues with an SVD

Kolmogorov n -width must be small ¹ $\mathcal{M}_h = \{u_h(\mu) \in V_h \mid \mu \in \mathcal{P}\}$ is a subset of a Banach space V_h .

The Kolmogorov n -width of \mathcal{M}_h in V_h is

$$d_n(\mathcal{M}_h, V_h) = \inf_{Y_n} \left\{ \sup_{x \in \mathcal{M}_h} \left(\inf_{y \in Y_n} \|x - y\|_{V_h} \right); Y_n \text{ is a } n\text{-dimensional subspace of } V_h \right\}. \quad (2)$$



¹A. Buffa, Y. Maday, A.T. Patera, C. Prudhomme, and G. Turinici, *A Priori convergence of the greedy algorithm for the parameterized reduced basis*.2010



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1 Compute the approximations $\{u_h(\mu_i)\}_{i=1,\dots,N}$.

2 We consider:

- A greedy algorithm with a Gram-Schmidt procedure
→ L^2 orthonormalization.
- Complemented by the following problem:
Find $\Phi^h \in X_h^N$, and $\lambda \in \mathbb{R}$ such that

$$\forall v \in X_h^N, \int_{\Omega} \nabla \Phi^h \cdot \nabla v = \lambda \int_{\Omega} \Phi^h \cdot v, \quad (3)$$

→ $L^2(\Omega)$ and $H^1(\Omega)$ orthogonalization.

$$X_h^N = \text{Vect}\{\Phi_1^h, \dots, \Phi_N^h\}$$



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3 Solve problem on the coarse mesh \mathcal{T}_H where $H \gg h$ with μ .

4 $\alpha_i^H = \int_{\Omega} I^h(u_H(\mu)) \cdot \Phi_i^h$ and **output**: $u_{Hh}^N = \sum_{i=1}^N \alpha_i^H \Phi_i^h$.

5 (Optional) Post-Treatment (PT)

$$\left\| u(x; \mu) - \sum_{k=1}^N (u_H(\mu), \phi_k^h) \phi_k^h \right\|_{H^1} \leq \underbrace{\epsilon}_{T_1} + \underbrace{C_1 h}_{T_2} + \underbrace{C_2 H^2}_{T_3} \sim \text{Ch}$$

if $H^2 \sim h$

where C_1, C_2 are constants independent of h and H .²

²Rachida Chakir, Yvon Maday. *A two-grid finite-element/reduced basis scheme for the approximation of the solution of parameter dependent PDE*. 2009

Results with FE solver



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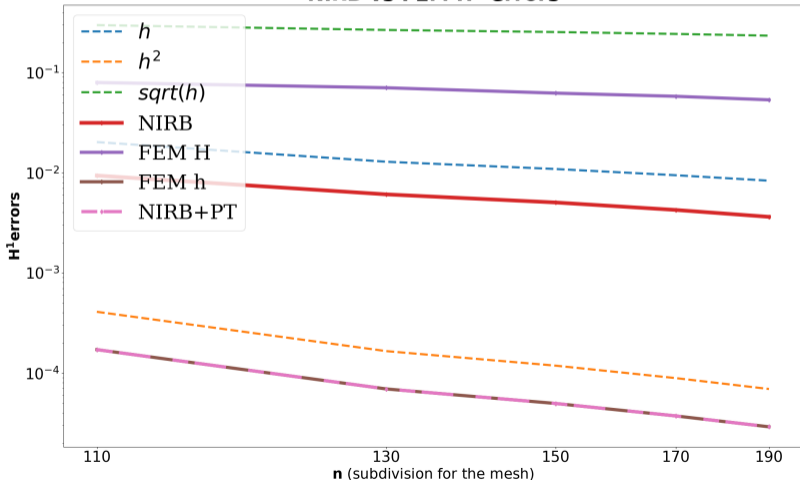
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NIRB vs FEM H^1 errors



Polytopal mesh for FV



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Goal:
Extend FE
estimate
to FV solver

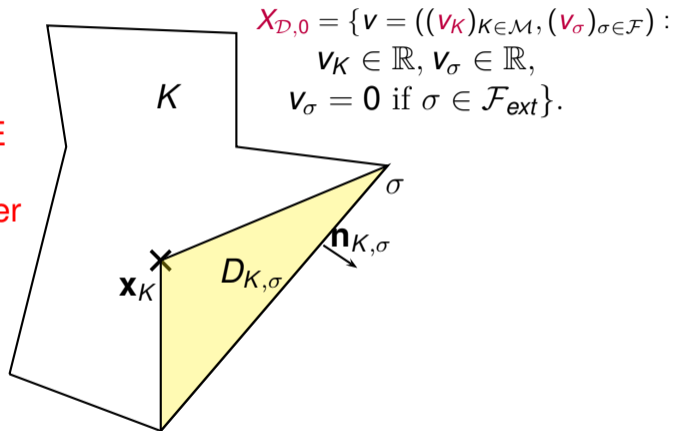


Figure: A cell K of a polytopal mesh ³

³J. Droniou, R. Eymard, T. Gallouët, C. Guichard, R. Herbin. *The gradient discretisation method*. 2018



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Variational Gradient Scheme ⁴

Find $u_{\mathcal{D}} \in X_{\mathcal{D},0}$ such that, $\forall v_{\mathcal{D}} \in X_{\mathcal{D},0}$,

$$\int_{\Omega} \mathbf{A}(\mu) \nabla_{\mathcal{D}} u_{\mathcal{D}} \cdot \nabla_{\mathcal{D}} v_{\mathcal{D}} = \int_{\Omega} f \Pi_{\mathcal{D}} v_{\mathcal{D}}. \quad (5)$$

⁴J. Droniou, R. Eymard, T. Gallouët, C. Guichard, R. Herbin. *The gradient discretisation method*. 2018

Error estimate on FV scheme



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1 $\Pi_{\mathcal{D}} : X_{\mathcal{D},0} \rightarrow L^2(\Omega) :$

$$\Pi_{\mathcal{D}} \mathbf{v}(\mathbf{x}) = v_K \text{ on } K.$$

2 $\nabla_{\mathcal{D}} : X_{\mathcal{D},0} \rightarrow L^2(\Omega)^d :$

$$\nabla_{\mathcal{D}} \mathbf{v}(\mathbf{x}) = \nabla_K \mathbf{v} + \mathbf{S} \text{ on } D_{K,\sigma}, \text{ where } \mathbf{S} \text{ ensures stability and}$$

$$\nabla_K \mathbf{v} = \frac{1}{|K|} \sum_{\sigma \in \mathcal{F}_K} |\sigma| v_{\sigma} \mathbf{n}_{K,\sigma}.$$

A norm on $X_{\mathcal{D},0}$: $\|\cdot\|_{\mathcal{D}} = \|\nabla_{\mathcal{D}} \cdot\|_{L^2(\Omega)^2}$.

H^1 error estimate

Main result: $\left\| u(\mu) - u_{Hh}^N(\mu) \right\|_{\mathcal{D}} \leq Ch, \text{ if } H^2 \sim h.$

Results with FV solver



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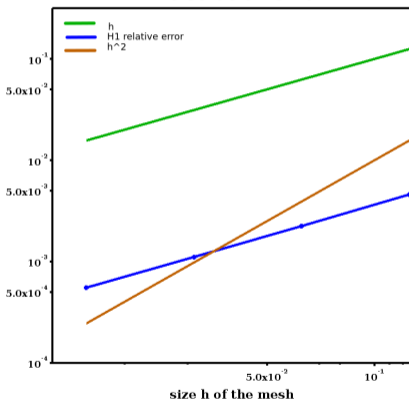
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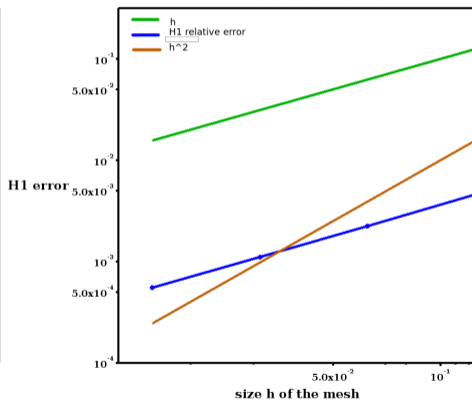
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Fine solution error



NIRB solution error



NIRB application: 2D Wind turbine



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Wind turbine

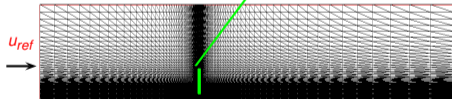


Figure 1: Mesh for one wind turbine

u_{ref} : Variable parameter

- 2D mesh with 6500 cells, thinner around the wind turbine.
- Characteristic length D : 126m, corresponds to the rotor diameter.
- Hub height: 95.6m.
- Wind turbine rotor is represented in the movement equation by adding a source term.
- **Boundary Condition:** u_{ref} at the inlet.
- Initial Condition: u_{ref} set in the domain.



Results for the application



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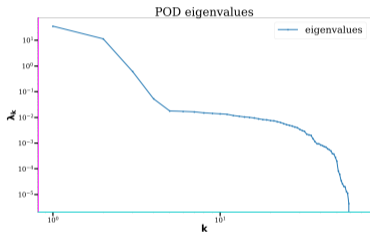


Figure 2: Decrease of the eigenvalues of the POD

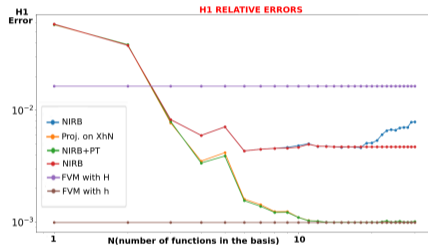


Figure 3: H^1 errors of the velocity on the region of interest



Wind turbines in 3D



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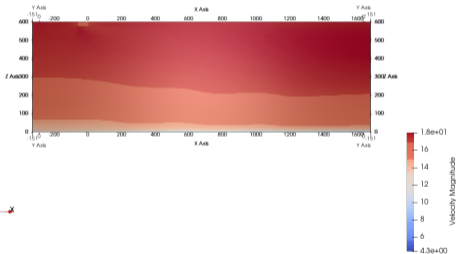
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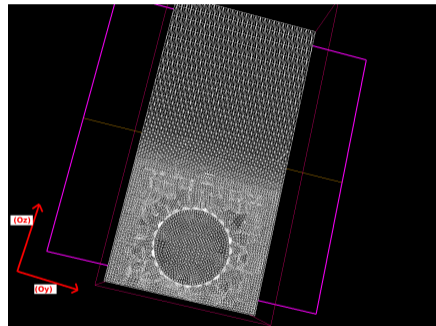
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Wind canal



One wind turbine mesh ($\mathcal{N} \sim 500\,000$)



Results for 3D application



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- Online
- Results with FE solver

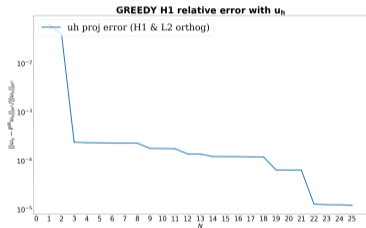
Finite volume solver

- Error estimate
- Results with FV solver

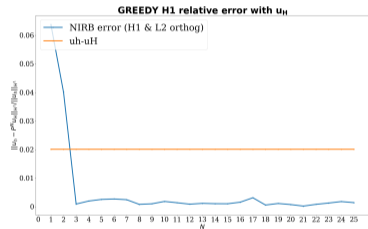
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Projection H^1 error



NIRB H^1 error

Wind turbine $n^{\circ}3$ approximation



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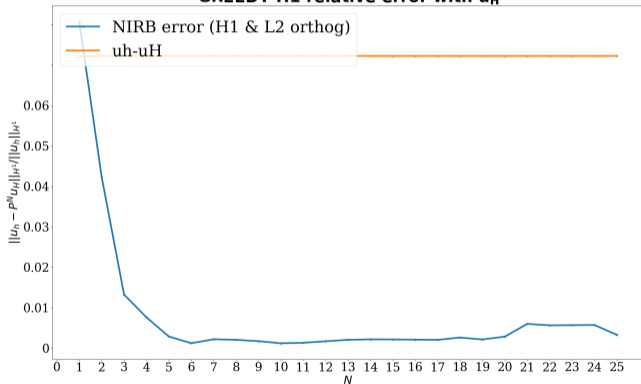
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We may approximate the third wind turbine with the second one!

GREEDY H1 relative error with u_H



NIRB H^1 error

Application on a truncated domain



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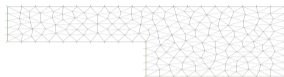
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Greedy on coarse mesh $(\Phi_i^H)_{i=1, \dots, N_1}$

Greedy on fine mesh $(\Phi_i^h)_{i=1, \dots, N_2}$

$$(A_i)_k = (u_H(\mu_k), \Phi_i^H)_{L^2}, \forall k = 1, \dots, N_{train}, \quad (6)$$

$$(B_i)_k = (u_h(\mu_k), \Phi_i^h)_{L^2}, \forall k = 1, \dots, N_{train}, \quad (7)$$

$$D = (A_1, \dots, A_N) \in \mathbb{R}^{N_{train} \times N_1}, \quad (8)$$

$$R_i = (D^T D + \lambda I_N)^{-1} D^T B_i, \quad \forall i = 1, \dots, N_2. \quad (9)$$

$$u_{Hh}^N(\mu) = \sum_{i=1}^{N_1} \sum_{j=1}^{N_2} R_{ij} (u_H(\mu), \Phi_j^H) \Phi_i^h. \quad (10)$$



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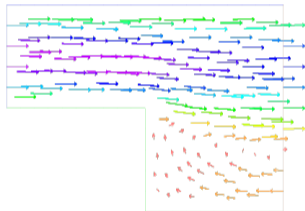
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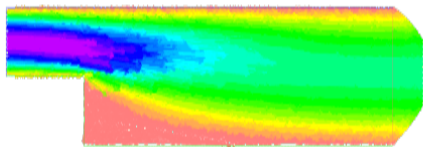
size of coarse mesh = 0.14

size of fine mesh = 0.03

H1 NIRB error for Reynolds = 200: $1.0549e-05$



FE solution on the coarse mesh



Nirb approximation



Conclusion and Perspectives



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- 1 Generalization of NIRB method to FV solvers with classical error estimate.
- 2 Numerical results with FV solver on wind turbines in accordance with expectations in 2D and 3D.
- 3 Efficient NIRB method on a truncated domain.

Perspectives

- Extend 3D wind turbines to offshore wind farm,
- Use a truncated domain on wind turbines,
- Generalize to other FV schemes,
- Development of a library of non intrusive reduced basis methods in process.



Thank you for your
attention!



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The rectification method

$$(u_H^i, \Phi_j) \rightarrow (u_h^i, \Phi_j)$$

$$(A_i)_k = (u_H(\mu_k), \Phi_i^h)_{L^2}, \forall k = 1, \dots, N_{train} \quad (11)$$

$$(B_i)_k = (u_h(\mu_k), \Phi_i^h)_{L^2}, \forall k = 1, \dots, N_{train} \quad (12)$$

$$D = (A_1, \dots, A_N) \in \mathbb{R}^{N_{train} \times N} \quad (13)$$

$$T_i = (D^T D + \lambda I_N)^{-1} D^T B_i, \forall i = 1, \dots, N. \quad (14)$$

$$u_{Hh}^N(\mu) = \sum_{i,j=1}^N T_{ij}(u_H(\mu), \Phi_j^h) \Phi_i^h \quad (15)$$