

Non Intrusive Reduced Basis method applied to wind turbine simulation

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Main goal

In an industrial context, parameter dependant PDEs may have to be solved numerically several times, and it can lead to very large computing times, for instance for:

- optimization paramaters fitting
- complex systems with uncertain scenarios

Non Intrusive Reduced Basis method (NIRB) is a quickly emerging field in applied mathematics and computational science. They enable high fidelity real-time simulations and **reduce the computational costs** thanks to the small Kolmogorov n-width [2]. These methods are decomposed in two stages: one **offline**, costly in time, and one **online**, much faster.

Industrial context

Code_Saturne is a free computer simulation software in fluid mechanics developed by EDF, using Finite Volumes method. One of the applications of Code_Saturne aims to obtain an optimal energy production by wind-farm. To do so, a specific velocity positioned upstream of a wind turbine is computed using a Reynolds Averaged Navier-Stokes (RANS) Equation for a given input velocity (uref). To further reduce calculation times, non intrusive reduced basis methods are suitable such as the two-grid method. RANS method gives mean quantities in time, and comes from the averaged Navier-Stokes equations:

$$\operatorname{div} \overline{\mathbf{u}} = 0, \tag{1}$$

$$\frac{\overline{u}_i}{\partial t} + \sum_{j=1}^{3} \overline{u}_j \frac{\partial \overline{u}_i}{\partial x_j} = \overline{f}_i - \frac{1}{\rho} \frac{\partial \overline{p}}{\partial x_i} + \sum_{j=1}^{3} \left[\nu \frac{\partial^2 \overline{u}_i}{\partial x_j \partial x_j} - \frac{\partial \overline{u}_i' \overline{u}_j'}{\partial x_j} \right], \tag{2}$$

deduced from the continuity and the momentum equations and the decomposition: $u_i = \overline{u}_i + u'_i$. Here, $\mathbf{u} = (u_1, u_2, u_3)$ is the velocity, ρ is the fluid density, p its pressure, ν the viscosity.



Two-grid method

The goal is to find a good approximation of the velocity on a fine mesh Ω_h using a velocity on a coarse mesh Ω_H .

To successfully apply reduced basis method, the Kolmogorov n-width must be very small. It means that the solution manifold $\mathcal{M}_h = \{\mathbf{u}_h(uref) \in X_h|uref \in \mathbb{R}^{Ntrain}\}$ for discrete solutions $\mathbf{u}_h(uref)$ of the RANS model on a fine mesh (of size h), may be approximated by a finite set of well-chosen solutions.

High-fidelity model (RANS)

Get snaphots

1. Run Code_Saturne as a **black-box** (no need to modify the code) to get discrete solutions for several parameters uref well chosen.

Output:
$$X = (\mathbf{u}_h^1, \dots, \mathbf{u}_h^{Ntrain}) \in \mathbb{R}^{\mathcal{N} \times Ntrain}$$

Greedy algorithm, POD, PGD,...

Build a reduced basis by POD

- **2.** Set the correlation Matrix $C_{i,j} = \int_{\Omega_h} \mathbf{u}_h^i \cdot \mathbf{u}_h^j$,
- **3.** Solve eigenvalues problem: $C\Psi_h^i = \lambda_i \Psi_h^i$, where $\Psi_h^i \in \mathbb{R}^{Ntrain}$.
- **4.** For N ordered proper values $(\lambda_1 > \cdots > \lambda_N)$, calculate N functions

$$\boldsymbol{\phi}_h^k = \sum_{j=1}^{Nrain} \boldsymbol{\Psi}_h^{k,j} \mathbf{u}_h^k \ \forall k = 1,...,N,$$

5. Normalize: $\boldsymbol{\phi}_h^k = \frac{\boldsymbol{\phi}_h^k}{\sqrt{\lambda_k}}$.

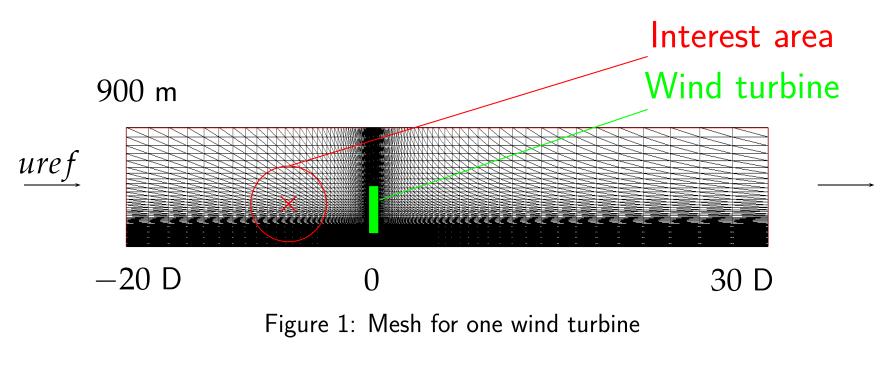
Output: $(\boldsymbol{\phi}_h^k)_{k=1,...N}$

NIRB

Reconstruct a fine approximation for a new parameter uref

6. A coarse mesh Ω_H is involved for the RANS resolution with H >> h the size of the mesh. Consequently, the time cost to obtain $\mathbf{u}_H(uref)$ is low. From the following estimates [1]: $\left\| \mathbf{u}(\mathbf{x}, uref) - \sum_{k=1}^{N} (\mathbf{u}_H(uref), \boldsymbol{\phi}_h^k)_{L^2} \boldsymbol{\phi}_h^k \right\|_{H^1} \leq C_1 h + C_2 H^2 + C(N) \varepsilon \sim o(h) \text{ if } H^2 \sim h,$ where C_1, C_2 , and C are constants independent of h and h, we deduce that $\mathbf{u}_H^N = \sum_{k=1}^{N} (\mathbf{u}_H(uref), \boldsymbol{\phi}_h^k)_{L^2} \boldsymbol{\phi}_h^k$ is a good approximation.

Application



Output: $\mathbf{u}_H^N \in X_h^N = Span\{\mathbf{u}_h^1, \dots, \mathbf{u}_h^N\}$

- ➤ 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.
- \blacktriangleright Boundary Condition: uref at the inlet.
- \blacktriangleright Initial Condition: uref set in the domain.

Results on one 2D wind turbine

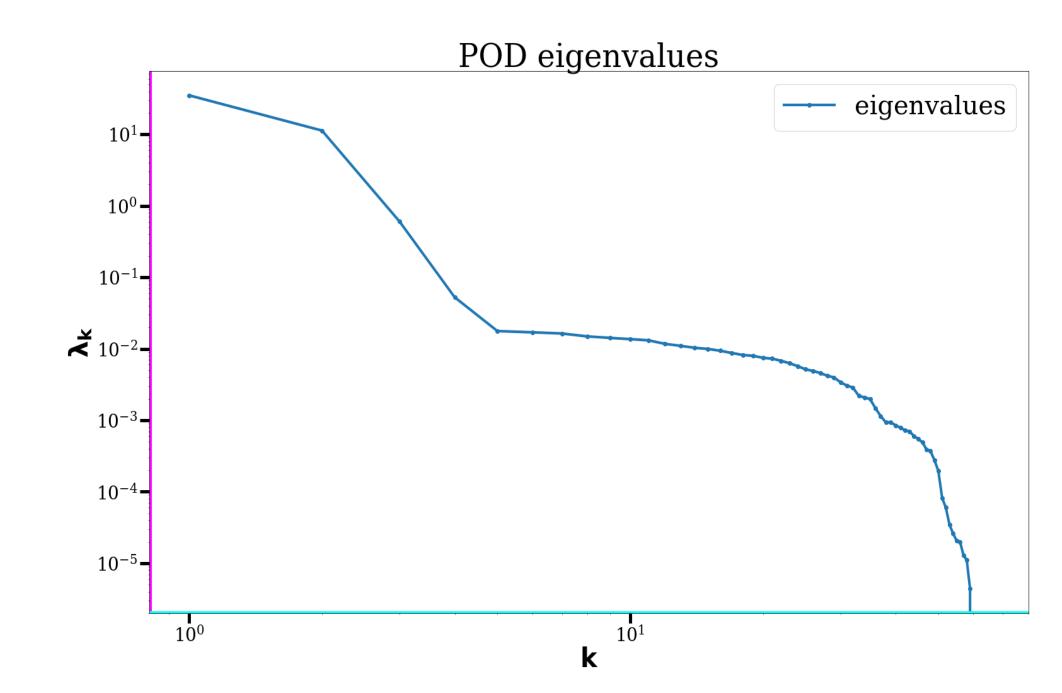


Figure 2: Decrease of the eigenvalues of the POD

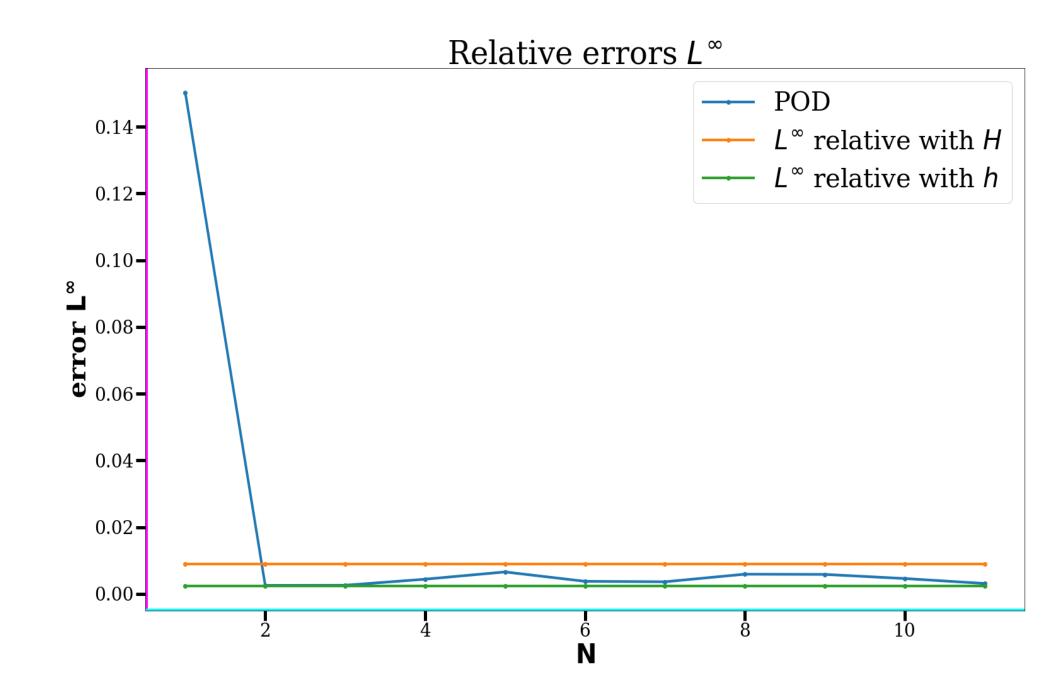


Figure 3: L^{∞} errors of the velocity on the interest area

- For k=3, $I(k)=\frac{\sum_{j=1}^k \lambda_j}{\sum_{j=1}^{Ntrain} \lambda_j}\simeq 1$, which means that the eigenvalues of POD decrease enough to perform NIRB method.
- After k = 5, the decay is slowed down (Figure 2), so the error of the POD increases slightly (Figure 3).
- The relative error of $\|\mathbf{u}_{h/10} \mathbf{u}_H^N\|_{L^{\infty}}$ is closer to the one given by $\|\mathbf{u}_{h/10} \mathbf{u}_h\|_{L^{\infty}}$ rather than the one with \mathbf{u}_H (Figure 3).

Conclusions and perspectives

- We deduce from these results that the two-grid method may be applied to this case of application.
- An analysis on the size H of the coarse mesh should provide the adequate size to obtain the best approximation.
- We plan to extend this method to the case of several wind-turbines.

References

- [1] R. Chakir, Y. Maday, A two-grid nite-element/reduced basis scheme for the approximation of the solution of parameter dependent P.D.E, 2009. hal-00387405f.
- [2] A. Buffa1, Y. Maday, A.T. Patera, C. Prudhomme and G. Turinici, *A priori convergence of the greedy algorithm for the parametrized reduced basis*, 2012. ESAIM: Mathematical Modelling and Numerical Analysis, EDP Sciences, 46 (3), pp.595-603.