

# The Reduced Basis method for aeroacoustic simulations solved by integral equations

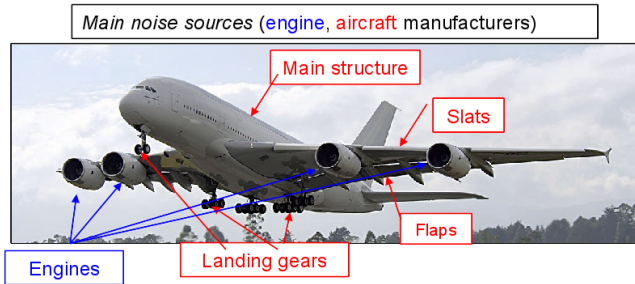
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joint work with Alexandre Ern and Tony Lelièvre

April 16, 2014

# Motivations

## Aeroacoustic in civil aviation



Numerical methods for computing acoustic scattering in civil aviation contexts

# Motivations

## Parametric problems

Different PDE's depending on physical assumptions, different parameters to take into account (position and frequency of the acoustic source, absorption coefficients, flow)

## Large scale computing

Uncertainty propagation, optimization, or real-time computation : need to solve parametric problem for a large number of parameter values

## Goal

Numerical procedure for fast computations of an approximation to the solution of a large class of parametric problem, with fast and accurate quantification of the approximation error

# Outline

## Two aeroacoustic problems solved by integral equations

Acoustic scattering by impedant objects in the air at rest

Acoustic scattering in moving air

## Some contributions to the Reduced Basis method

(A quick reminder on) The reduced basis method

An accurate and online-efficient evaluation of the error bound

A nonintrusive technique for the RB method

## Some numerical illustrations

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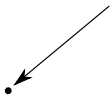
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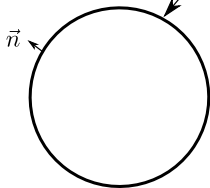
## Some numerical illustrations

# Acoustic scattering by impedant objects in the air at rest

acoustic monopole



impedant surface  $\Gamma$



$\vec{n}$

## Acoustics problem

### Variational formulation

Find  $\chi, \lambda : \Gamma \rightarrow \mathbb{C}$  s.t. for all  $\chi^t, \lambda^t$

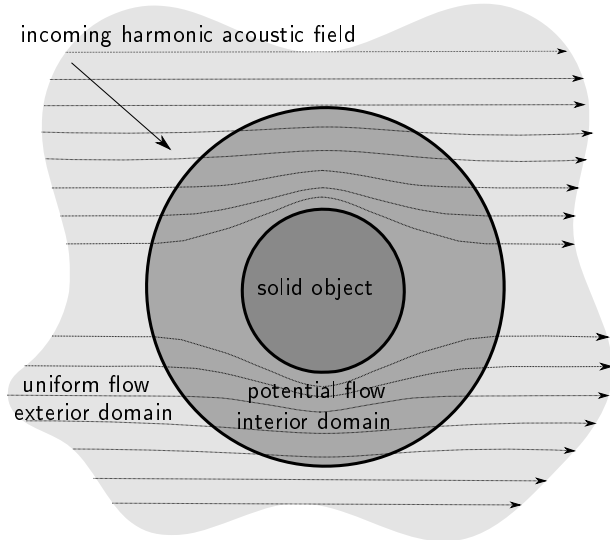
$$\begin{cases} \left( N\chi - \frac{ik}{2\mu}\chi, \chi^t \right) + \left( \tilde{D}\lambda, \chi^t \right) = (\gamma_1 p_{\text{inc}}, \chi^t), \\ \left( D\chi, \lambda^t \right) - \left( S\lambda + \frac{i\mu}{2k}\lambda, \lambda^t \right) = -(\gamma_0 p_{\text{inc}}, \lambda^t), \end{cases}$$

- $\mu$  is the impedant coefficient, quantifies the absorbed and scattered parts of the acoustic pressure field
- $N, D, \tilde{D}$  and  $S$  are integral operators involving the Green kernel associated to the Helmholtz equation

$$G(x, y) = \frac{\exp(ik|x - y|)}{4\pi|x - y|}$$

- $p_{\text{inc}}(\mathbf{x}) = A \frac{\exp(ik|x - x_S|)}{4\pi|x - x_S|}$  for a monopole source of amplitude  $A$  located at  $x_S$
- $k = \frac{2\pi f}{c}$ , with  $f$  the frequency of the source

# Acoustic scattering in moving air





# Aeroacoustics problem

## Characteristics

- acoustics in a mean flow not at rest
- unbounded domain of propagation

## Tools

- Linearization of the Euler equations → convected Helmholtz equation
- Prandtl–Glauert transformation [Glauert (1928), Amiet and Sears (1970)]  
→ transforms the uniformly convected Helmholtz eq. into the classical Helmholtz eq. in the exterior domain
- Coupled BEM-FEM [McDonald and Wexler (1972), Zienkiewicz, Kelly and Bettess (1977), Johnson and Nédélec (1980)]
- Stabilization with respect to resonant frequencies [Buffa and Hiptmair (2005), Hiptmair and Meury (2006)]

## Variational formulation

$$\left\{ \begin{array}{l} \mathcal{V}(f, f^t) + (N(\gamma_0 f), \gamma_0 f^t)_{\Gamma_\infty} + \left( \left( \tilde{D} - \frac{1}{2}I \right) (\lambda), \gamma_0 f^t \right)_{\Gamma_\infty} = (\gamma_1 f_{\text{inc}}, \gamma_0 f^t)_{\Gamma_\infty} \\ \left( \left( D - \frac{1}{2}I \right) (\gamma_0 f), \lambda^t \right)_{\Gamma_\infty} - (S(\lambda), \lambda^t)_{\Gamma_\infty} + i(p, \lambda^t)_{\Gamma_\infty} = -(\gamma_0 f_{\text{inc}}, \lambda^t)_{\Gamma_\infty} \\ (N(\gamma_0 f), p^t)_{\Gamma_\infty} + \left( \left( \tilde{D} + \frac{1}{2}I \right) (\lambda), p^t \right)_{\Gamma_\infty} - (p, p^t)_{H^1(\Gamma_\infty)} = (\gamma_1 f_{\text{inc}}, p^t)_{\Gamma_\infty} \end{array} \right.$$

- $\mathcal{V}(f, f^t) = \int_{\Omega^-} r \Xi \nabla \bar{f} \cdot \nabla f^t - \int_{\Omega^-} r k^2 \beta \bar{f} f^t + i \int_{\Omega^-} r k \mathbf{V} \cdot (\bar{f} \nabla f^t - f^t \nabla \bar{f})$
- $N$ ,  $D$ ,  $\tilde{D}$  and  $S$  are integral operators
- $f_{\text{inc}}(\mathbf{x}) = A \frac{\exp(ik|\mathbf{x} - \mathbf{x}_S|)}{4\pi|\mathbf{x} - \mathbf{x}_S|}$  for a monopole source of amplitude  $A$  located at  $\mathbf{x}_S$
- $k = \frac{2\pi f}{c}$ , with  $f$  the frequency of the source

## Well-posed formulation

### Fredholm alternative

- uniqueness : Rellich lemma and unique continuation property, with “weakly regular” coefficients [Garofalo et Lin (1987)]
- isolate a coercive part of the sesquilinear form (including stabilization)
- compact perturbation

### Finite dimension approximation

- tetrahedral mesh, meshsize  $h$
- finite elements P1 for  $f$  and  $p$ , and finite elements P0 for  $\lambda$
- inf-sup stable discrete formulation (for  $h$  small enough)

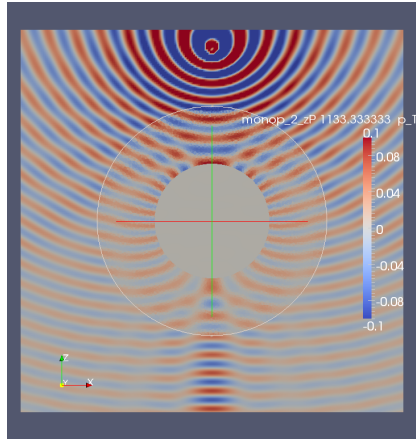
### Conclusion

- Direct use of BEM codes for Helmholtz
- Numerical method valid for all frequencies of the source

### Well-posed variational formulation

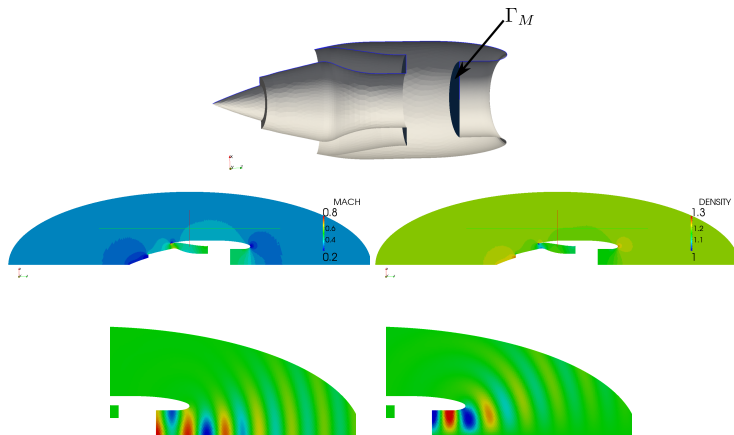
[F.C., A. Ern and G. Sylvand (2014)]

## Numerical illustration



Computation by Airbus (350k dof, FMM : 1h30 sur 32 proc)

# Acoustic scattering by a turbojet



Computation by Airbus (4.7M ddl, FMM : 1h30 sur 160 proc)

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# The reduced basis method

## Parametrized problem - idealized case

$$a_\mu(u_\mu, v) = b(v) \quad \forall v \in \mathcal{V}$$

where

- $\mu \in \mathcal{P}$  is the parameter
- $a_\mu$  is a continuous and coercive bilinear form (uniformly in  $\mu$ )
- $b$  is a continuous linear form
- $\mathcal{V}$  is a  $N$ -dimensional linear space, approx. of a Hilbert space  $H$ ,  $N \gg 1$

## Goal

Compute an approximation  $\hat{u}_\mu$  of  $u_\mu$  and (sharp) upper bound of  $|\hat{u}_\mu - u_\mu|$  in complexity independent of  $N$

## Some early references

[Machiels, Maday, Oliveira, Patera and Rovas (2000)]

[Prud'homme, Rovas, Veroy, Machiels, Maday, Patera and Turinici (2002)]

# Principles

## Offline stage

- some parameter values  $(\mu_i)_{1 \leq i \leq \hat{N}}$  are chosen in  $\mathcal{P}_{\text{trial}} \subset \mathcal{P}$  by a greedy algorithm,  $\hat{N} \ll N$
- the full solutions  $u_{\mu_i}$  are computed for all  $1 \leq i \leq \hat{N}$
- approximation space  $\hat{\mathcal{V}}_{\hat{N}} = \text{Vect}\{u_{\mu_1}, \dots, u_{\mu_{\hat{N}}}\}$

## Online problem

Galerkin procedure on  $\hat{\mathcal{V}}_{\hat{N}}$

$$a_{\mu}(\hat{u}_{\mu}, u_{\mu_j}) = b(u_{\mu_j}) \quad \forall j \in \{1, \dots, \hat{N}\}$$

The decomposition of  $\hat{u}_{\mu}$  in  $\hat{\mathcal{V}}_{\hat{N}}$  is denoted

$$\hat{u}_{\mu} = \sum_{i=1}^{\hat{N}} \gamma_i(\mu) u_{\mu_i}$$



## A posteriori error bound

### Residual

Let  $G_\mu : H \rightarrow H$  such that

$$\langle G_\mu u, v \rangle_H = a_\mu(u, v) - b(v)$$

where  $\langle \cdot, \cdot \rangle_H$  is the inner product of  $H$

### A posteriori error bound

$$\mathcal{E}(\mu) = \frac{\|G_\mu \hat{u}_\mu\|_H}{\beta} \geq \|\hat{u}_\mu - u_\mu\|_H$$

where  $\beta$  is the coercivity constant of  $a_\mu$  and  $\|u\|_H = \sqrt{\langle u, u \rangle_H}$

### Goal-oriented

Need to introduce a dual problem

## Online-efficiency

### Affine dependence

$a_\mu(\cdot, \cdot)$  has an affine dependence in  $\mu$  if there exists  $\mu \mapsto \alpha_k(\mu)$  and  $a_k(\cdot, \cdot)$ ,  $1 \leq k \leq d$ , such that  $a_\mu(\cdot, \cdot) = \sum_{k=1}^d \alpha_k(\mu) a_k(\cdot, \cdot)$

### The key assumption

**Affine dependence** allows to construct the online problem and compute the error bound in **complexity independent of  $N$** , since

$$\begin{aligned} a_\mu(u_{\mu_i}, u_{\mu_j}) &= \sum_{k=1}^d \alpha_k(\mu) a_k(u_{\mu_i}, u_{\mu_j}) \\ \mathcal{E}(\mu) &= \frac{1}{\beta} \left( \langle Jb, Jb \rangle_H - 2 \sum_{k=1}^d \sum_{i=1}^{\hat{N}} \alpha_k(\mu) \gamma_i(\mu) \langle Jb, Ja_k(u_{\mu_i}, \cdot) \rangle_H \right. \\ &\quad \left. + \sum_{k=1}^d \sum_{l=1}^d \sum_{i=1}^{\hat{N}} \sum_{j=1}^{\hat{N}} \alpha_k(\mu) \gamma_i(\mu) \alpha_l(\mu) \gamma_j(\mu) \langle Ja_k(u_{\mu_i}, \cdot), Ja_l(u_{\mu_j}, \cdot) \rangle_H \right)^{\frac{1}{2}} \end{aligned}$$

where the only  $N$ -dependent quantities are  $\mu$ -independent, and therefore can be **precomputed** during the offline stage

## Two bottlenecks are investigated

1 - the formula for the error bound is very sensitive to round-off errors

Denote

- $\mathcal{E}_1(\mu) = \frac{\|G_\mu \hat{u}_\mu\|_H}{\beta}$

- $\mathcal{E}_2(\mu)$  = online-efficient formula

$\mathcal{E}_1(\mu) = \mathcal{E}_2(\mu)$  in exact arithmetics

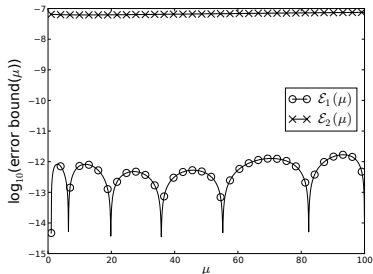
Simple example

$$-u'' + \mu u = 1$$

on  $]0, 1[$  with  $u(0) = u(1) = 0$

2 - the RB method is intrusive

The **precomputations** in  $a_\mu(u_{\mu_i}, u_{\mu_j}) = \sum_{k=1}^d \alpha_k(\mu) a_k(u_{\mu_i}, u_{\mu_j})$  suppose that we can modify the elementary assembly routines of the code



## New formula for computing the error bound

### Rewriting of $\mathcal{E}_2$

Define

- $x(\mu) \in \mathbb{R}^{d\hat{N}}$ , with components  $\alpha_k(\mu)\gamma_i(\mu)$ ,  $1 \leq k \leq d$ ,  $1 \leq i \leq \hat{N}$
  - $\sigma = 1 + d\hat{N} + \frac{1}{2}d\hat{N}(d\hat{N} + 1)$
  - $X(\mu) \in \mathbb{R}^\sigma$ , with components  $(1, x_l(\mu), x_l(\mu)x_J(\mu))$ ,  $1 \leq l \leq J \leq d\hat{N}$
- $(\beta\mathcal{E}(\mu))^2$  is a linear form in  $X(\mu) : \exists q \in \mathbb{R}^\sigma$ , independent of  $\mu$ , s.t.

$$(\beta\mathcal{E}(\mu))^2 = \sum_{p=1}^{\sigma} q_p X_p(\mu)$$

### Separated representation of $X_p(\mu)$

Empirical Interpolation Method (EIM) [Barrault, Maday, Nguyen, Patera (2004)]

## Empirical Interpolation Method for $X_p(\mu)$

- Initialization  $k = 1$ 
  - Compute  $p_1 := \operatorname{argmax}_{p \in \{1, \dots, s\}} \|X_p(\cdot)\|_{\ell^\infty(\mathcal{P}_{\text{trial}})}$
  - Compute  $\mu_1 := \operatorname{argmax}_{\mu \in \mathcal{P}_{\text{trial}}} |X_{p_1}(\mu)|$  and set  $\mathcal{P}_{\text{inter}} = \{\mu_1\}$
  - Set  $q_1(\cdot) := \frac{X_{p_1}(\cdot)}{X_{p_1}(\mu_1)}$  and  $B_{11}^1 := 1$
- While  $k < \hat{\sigma}$ 
  - Compute  $p_{k+1} := \operatorname{argmax}_{p \in \{1, \dots, s\}} \|(\delta^k X)_p(\cdot)\|_{\ell^\infty(\mathcal{P}_{\text{trial}})}$
  - Compute  $\mu_{k+1} := \operatorname{argmax}_{\mu \in \mathcal{P}_{\text{trial}}} |(\delta^k X)_{p_{k+1}}(\mu)|$  and set  $\mathcal{P}_{\text{inter}} := \mathcal{P}_{\text{inter}} \cup \{\mu_{k+1}\}$
  - Set  $q_{k+1}(\cdot) := \frac{(\delta^k X)_{p_{k+1}}(\cdot)}{(\delta^k X)_{p_{k+1}}(\mu_{k+1})}$  and  $B_{ij}^{k+1} := q_j(\mu_i)$ ,  $1 \leq i, j \leq k+1$
  - $k \leftarrow k+1$

where  $\delta^k := \operatorname{Id} - I^k$  and

$$I^k X(\mu) := \sum_{r=1}^k \lambda_r^k(\mu) X(\mu_r),$$

with  $\lambda^k(\mu) \in \mathbb{C}^k$  solves  $B^k \lambda^k(\mu) = q^k(\mu)$

## Empirical Interpolation Method for $X_p(\mu)$

### Change of base

Since  $\text{Vect}_{1 \leq m \leq k}(q_m(\cdot)) = \text{Vect}_{1 \leq m \leq k}(X_{p_m}(\cdot))$ , there exists a matrix  $\Gamma \in \mathbb{R}^{k \times k}$  such that, for all  $1 \leq l \leq k$ ,

$$\sum_{m=1}^k \Gamma_{lm} q_m(\mu) = X_{p_m}(\mu), \quad \forall \mu \in \mathcal{P}$$

$\Gamma$  is constructed recursively :

- $k = 1$  :

$$\Gamma_{1,1} = X_{p_1}(\mu_1),$$

- $k \rightarrow k + 1$  :

$$\Gamma_{k+1,k+1} = (\delta_k X)_{p_{k+1}}(\mu_{k+1}),$$

$$\Gamma_{l,k+1} = 0, \quad \forall 1 \leq l \leq k,$$

$$\Gamma_{k+1,l} = \kappa_l, \quad \forall 1 \leq l \leq k,$$

where  $\kappa$  is such that  $\sum_{m=1}^k B_{lm} \kappa_m = X_{p_{k+1}}(\mu_l)$ , for all  $1 \leq l \leq k$

## New formula for computing the error bound

Separated representation of  $X_p(\mu)$

$$X_p(\mu) \approx (I^{\hat{\sigma}} X)_p(\mu) = \sum_{i=1}^{\hat{\sigma}} \sum_{j=1}^{\hat{\sigma}} \Delta_{ij} X_{p_j}(\mu) X_p(\mu_i) \quad \hat{\sigma} \leq \sigma$$

where  $\Delta := B^{-t} \Delta^{-1}$

New online estimator

Using the separated representation of  $X(\mu)$  and exchanging summation order

$$\begin{aligned} (\beta \mathcal{E}(\mu))^2 &\approx \sum_{p=1}^{\sigma} q_p \left\{ \sum_{i=1}^{\hat{\sigma}} \sum_{j=1}^{\hat{\sigma}} \Delta_{ij} X_{p_j}(\mu) X_p(\mu_i) \right\} \\ &= \sum_{i=1}^{\hat{\sigma}} \underbrace{\left\{ \sum_{j=1}^{\hat{\sigma}} \Delta_{ij} X_{p_j}(\mu) \right\}}_{\lambda_i(\mu)} \underbrace{\left\{ \sum_{p=1}^{\sigma} q_p X_p(\mu_i) \right\}}_{(\beta \mathcal{E}_1(\mu_i))^2} \end{aligned}$$

leading to

$$\mathcal{E}_3(\mu) := \frac{1}{\beta} \left( \sum_{i=1}^{\hat{\sigma}} \lambda_i(\mu) (\beta \mathcal{E}_1(\mu_i))^2 \right)^{\frac{1}{2}}$$

## New formula for computing the error bound

### An online-efficient formula

The accurate evaluations  $\mathcal{E}_1(\mu_i)$ ,  $1 \leq i \leq \hat{\sigma}$ , are precomputed during the offline stage and  $\lambda_i(\mu)$  are computed in complexity independent of  $N$

### Summary

	$\mathcal{E}_1(\mu)$	$\mathcal{E}_2(\mu)$	$\mathcal{E}_3(\mu)$
Accuracy scales like	$\epsilon$	$\sqrt{\epsilon}$	$\epsilon$
Online efficiency	No	Yes	Yes

where  $\epsilon$  is the machine precision

### Usefulness

- bad stability constant (can be the case for the Helmholtz equation)
- nonlinear problems. Brezzi-Rappaz-Raviart theory : no error bound is possible until a very tight tolerance is reached (“overkill”).  $\mathcal{E}_3$  used in [M. Yano (2013)] for Boussinesq certified RB at Grashof numbers of engineering interest.

### References

[F.C. (2012)]

[F.C., A. Ern and T. Lelièvre (2013)]



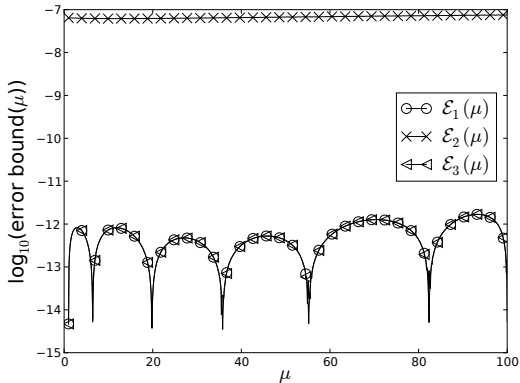
## Illustration on a simple case

Back to the simple problem

$$-u'' + \mu u = 1$$

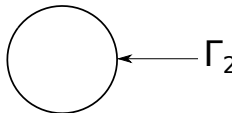
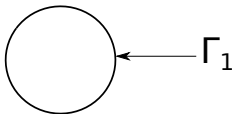
on  $]0, 1[$  with  $u(0) = u(1) = 0$

- $d = 2, \hat{N} = 7$   
 $\implies \sigma = 120$
- $\hat{\sigma} = 23$



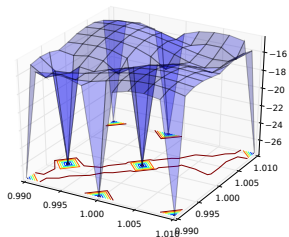
## Acoustics problem

acoustic monopole

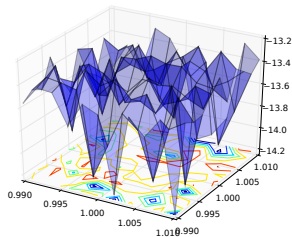


$$\begin{cases} \left( N\chi - \frac{ik}{2\mu}\chi, \chi^t \right) + \left( \tilde{D}\lambda, \lambda^t \right) = (\gamma_1 p_{\text{inc}}, \chi^t), \\ \left( D\chi, \lambda^t \right) - \left( S\lambda + \frac{i\mu}{2k}\lambda, \lambda^t \right) = -(\gamma_0 p_{\text{inc}}, \lambda^t), \end{cases}$$

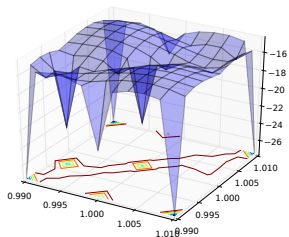
# Acoustics problem



$\mathcal{E}_1$



$\mathcal{E}_2$



$\mathcal{E}_3$

Two scattering balls, s.t.

- $d = 5$ ,  $\hat{N} = 8$   
 $\implies \sigma = 1681$  (complex)
- $\hat{\sigma} = 60$

# Nonintrusivity of reduced problem construction

## Example of affine decomposition

$$A_{\mu,ij} = \underbrace{\int_{\Omega} \nabla \varphi_i \cdot \nabla \varphi_j dx}_{:=A_0 ij} + \mu \underbrace{\int_{\Omega} \varphi_i \varphi_j dx}_{:=A_1 ij}, \quad 1 \leq i, j \leq N$$

Computing  $A_0$  and  $A_1$  requires entering the assembly routines of the code

- not feasible for large industrial codes
- **nonintrusivity** : use only full matrix  $A_{\mu}$  at user-selected  $\mu$ 's

## Simple fix

Select two values of the parameter and observe that

$$A_{\mu} = \frac{\mu_2 - \mu}{\mu_2 - \mu_1} A_{\mu_1} + \frac{\mu - \mu_1}{\mu_2 - \mu_1} A_{\mu_2}$$

## Goal 1

Extend this idea for more complex parameter dependencies

## From affine dependence to nonintrusivity

### General affine dependence

$$A_{\mu,ij} = \sum_{s=1}^{\varsigma} g_s(\mu) \int_{\Omega} \psi_{s,ij}(x) dx$$

(previous example :  $\varsigma = 2$ ,  $g_1(\mu) = 1$ ,  $\psi_{1,ij}(x) = \nabla \varphi_i(x) \cdot \nabla \varphi_j(x)$ ,  $g_2(\mu) = \mu$ ,  $\psi_{2,ij}(x) = \varphi_i(x)\varphi_j(x)$ )

### Key idea

Separate  $\mu$  and  $s$  in  $g_s(\mu)$  using EIM (of rank  $d \leq \varsigma$ )

$$g_s(\mu) \approx (I_d g)(\mu, s) = \sum_{k=1}^d \underbrace{\left\{ \sum_{l=1}^d \Delta_{kl} g_{s_l}(\mu) \right\}}_{:=\beta_k(\mu)} g_s(\mu_k)$$

### Nonintrusive approximation

Exchange summation order

$$A_{\mu,ij} \approx \sum_{s=1}^{\varsigma} (I_d g)(\mu, s) \int_{\Omega} \psi_{s,ij}(x) dx = \sum_{k=1}^d \beta_k(\mu) A_{\mu_k,ij}$$

## Nonaffine parameter dependencies

### General case

$$A_{\mu,ij} = \sum_{s=1}^{\varsigma} \int_{\Omega} g_s(\mu, x) \psi_{s,ij}(x) dx$$

### Goal 2

Achieve nonintrusivity in this general setting

### Step 1 - classical

Use  $\varsigma$  EIM's (same rank  $d$  for simplicity) to separate  $(\mu, x)$  in the  $g_s$ 's :  
Approximate  $A_{\mu,ij} \approx I_d^g A_{\mu,ij}$  with

$$I_d^g A_{\mu,ij} = \sum_{s=1}^{\varsigma} \sum_{k=1}^d \underbrace{\left\{ \sum_{l=1}^d \Delta_{s,kl} g_s(\mu, x_{s,m}) \right\}}_{:=z_{sk}(\mu)} \underbrace{\int_{\Omega} g_s(\mu_{s,k}, x) \psi_{s,ij}(x) dx}_{:=Q_{sk,ij}}$$

## Nonaffine parameter dependencies

### Regroup indices

s and k to write ( $d^p = \varsigma d$ )

$$I_d^g A_{\mu,ij} = \sum_{p=1}^{d^p} z_p(\mu) Q_{p,ij}$$

### Step 2 - key idea

Separate  $(\mu, p)$  in  $z_p(\mu)$  using a second EIM (rank  $d^z \leq d^p$ )

$$z_p(\mu) \approx \sum_{k=1}^{d^z} \sum_{l=1}^{d^z} \Delta_{kl}^{(z)} z_{p_l}^{(\mu)}(\mu) z_p(\mu_k^{(\mu)})$$

## Nonaffine parameter dependencies

### Nonintrusive formula

Exchanging summation order

$$\begin{aligned}
 A_{\mu,ij} &\approx I_d^g A_{\mu,ij} = \sum_{p=1}^{d^P} \left\{ \sum_{k=1}^{d^Z} \sum_{l=1}^{d^Z} \Delta_{kl}^{(z)} z_{p_l}^{(\mu)}(\mu) z_p(\mu_k^{(z)}) \right\} Q_{p,ij} \\
 &= \sum_{k=1}^{d^Z} \underbrace{\left\{ \sum_{l=1}^{d^Z} \Delta_{kl}^{(z)} z_{p_l}^{(z)}(\mu) \right\}}_{:=\beta_k^{(z)}(\mu)} \underbrace{\left\{ \sum_{p=1}^{d^P} z_p(\mu_k^{(z)}) Q_{p,ij} \right\}}_{=I_d^g A_{\mu_k^{(z)},ij} \approx A_{\mu_k^{(z)},ij}} \\
 &\approx \sum_{k=1}^{d^Z} \beta_k^{(z)}(\mu) A_{\mu_k^{(z)},ij}
 \end{aligned}$$



## “Nonintrusive” Reduced Basis

### Not black-box

Coefficients of the reduced matrix

$$\hat{A}_{\mu,ij} = \sum_{k=1}^{d^z} \beta_k^{(z)}(\mu) \hat{U}_i^t A_{\mu_k^{(z)}} \hat{U}_j$$

Nonintrusive in the sense that we only need

- the function  $(\mu, V) \mapsto A_{\mu} V$  (quite a mild assumption). Any optimized matrix-vector product available in the industrial code can be directly used
- to know the terms of the variational formulation

### Reference

[F.C., A. Ern and T. Lelièvre], submitted

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Acoustic scattering in moving air

## Some contributions to the Reduced Basis method

(A quick reminder on) The reduced basis method

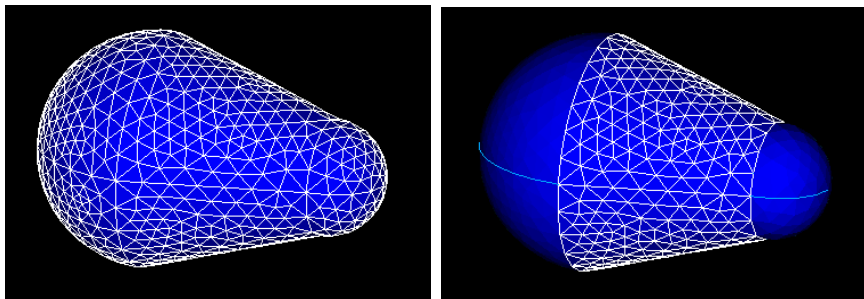
An accurate and online-efficient evaluation of the error bound

A nonintrusive technique for the RB method

## Some numerical illustrations

# Acoustic scattering by impedant objects in the air at rest

Mesh, 2240 dof



## Parametric problem

- parameters : frequency of the source (plane wave), 3 impedant coefficients
- quantity of interest : far-field acoustic scattered field on the axis of symmetry

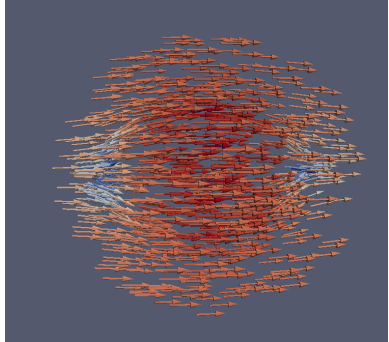
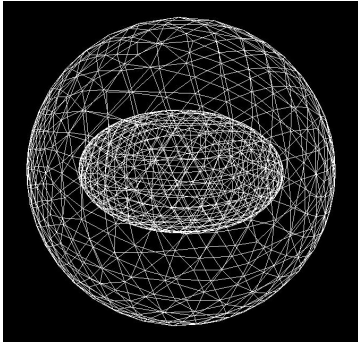
## Acoustic scattering by impedant objects in the air at rest

### Reduced basis

- approximation formula for the matrix : 20 terms, for the right-hand side of the direct and dual problems : 13 terms
- 20 basis vectors (max of the error bound  $10^{-6}$ )
- speed-up factor  $> 10^4$  ( 2.8 ms vs 30 s)

# Acoustic scattering in moving air

Mesh, 1711 dof



## Parametric problem

- parameters : frequency of the source (monopole), uniform perturbation of the flow (3 components)
- quantity of interest : acoustic potential at a point

## Acoustic scattering in moving air

Frequency-dependent terms

Flow-dependent terms

$$\left\{ \begin{array}{l} \mathcal{V}(f, f^t) + (N(\gamma_0 f), \gamma_0 f^t)_{\Gamma_\infty} + \left( \left( \tilde{D} - \frac{1}{2}I \right) (\lambda), \gamma_0 f^t \right)_{\Gamma_\infty} = (\gamma_1 f_{\text{inc}}, \gamma_0 f^t)_{\Gamma_\infty} \\ \left( \left( D - \frac{1}{2}I \right) (\gamma_0 f), \lambda^t \right)_{\Gamma_\infty} - (S(\lambda), \lambda^t)_{\Gamma_\infty} + i(p, \lambda^t)_{\Gamma_\infty} = -(\gamma_0 f_{\text{inc}}, \lambda^t)_{\Gamma_\infty} \\ (N(\gamma_0 f), p^t)_{\Gamma_\infty} + \left( \left( \tilde{D} + \frac{1}{2}I \right) (\lambda), p^t \right)_{\Gamma_\infty} - (p, p^t)_{H^1(\Gamma_\infty)} = (\gamma_1 f_{\text{inc}}, p^t)_{\Gamma_\infty} \end{array} \right.$$

with  $\mathcal{V}(f, f^t) = \int_{\Omega^-} r \bar{\mathbf{E}} \nabla \bar{f} \cdot \nabla f^t - \int_{\Omega^-} r k^2 \beta \bar{f} f^t + i \int_{\Omega^-} r k \mathbf{V} \cdot (\bar{f} \nabla f^t - f^t \nabla \bar{f})$

## Acoustic scattering in moving air

### Reduced basis

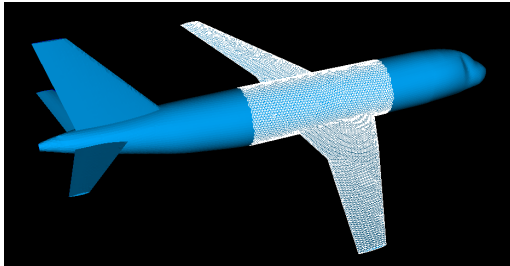
- approximation formula for the matrix : 25 terms, for the right-hand side of the direct and dual problems : 18 terms
- 20 basis vectors (max of the error bound  $10^{-7}$ )
- speed-up factor  $> 5 \times 10^3$  ( 2.8 ms vs 14 s)

## Towards industrial applications

### “Scalable” implementation

- never save a matrix on hard-drive
- fast matrix-vector products (parallel FMM)
- parallel exploration of  $\mathcal{P}_{\text{trial}}$

Mesh, 60866 dof



parameters : frequency of the source (monopole), 3 impedant coefficients



## Towards industrial applications

### Reduced basis

- approximation formula for the matrix : 50 terms, for the right-hand side of the direct problem : 60 terms
- 30 basis vectors
- offline stage :  $\approx 2$  days, exploration of  $\mathcal{P}_{\text{trial}} \approx 1$  hour at the 30th loop
- speed-up factor  $\approx 1.6 \times 10^5$  ( 15 ms vs 40 min)
- all computations of this laptop with 4 CPUs and 4 GB of RAM

# Towards industrial applications

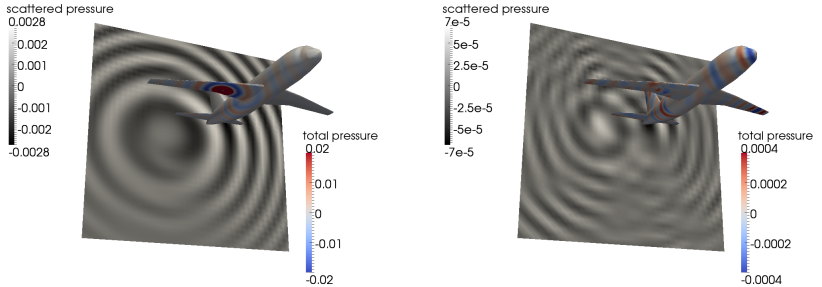


Figure: Left : direct solution, right : difference between direct and RB solutions

Real-time online computation

⇒ demonstration