

# Model Order Reduction Techniques I & II

## RB: General Preliminaries, Elliptic Problems

M. Grepl<sup>a</sup> & K. Veroy-Grepl<sup>b</sup>

<sup>a</sup>Institut für Geometrie und Praktische Mathematik

<sup>b</sup>High Performance Computation for Engineered Systems

RWTH Aachen

Sommersemester 2019

## General Preliminaries

The Poisson Problem

Inner Product Spaces

Linear and Bilinear Forms

Classes of Functions

Scalar and Vector Fields

Function Spaces

## Linear Elliptic Problems I

Problem Statement

Truth Approximation

## Reduced Basis Approximation

Objective

RB Approximation Space

Galerkin Projection

Offline-Online Computational Procedure

## Strong Formulation – Dirichlet Problem

Find  $u$  such that

$$\begin{array}{ll} -\nabla^2 u = f & \text{in } \Omega \\ u = 0 & \text{on } \Gamma \end{array}$$

where

$$\nabla^2 \equiv \frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2}$$

and  $\Omega$  is a domain in  $\mathbf{R}^2$  with boundary  $\Gamma$ .

In general, we assume

- ▶  $\Omega \subset \mathbf{R}^d$ ,  $d = 1, \dots, 3$ , has a Lipschitz continuous boundary  $\Gamma$  (or  $\partial\Omega$ )
- ▶  $\Omega$  is an open domain, e.g.,  $\Omega = (0, 1)$  ( $\bar{\Omega} = [0, 1]$ )

## Minimization Principle – Statement

Find

$$u = \arg \min_{w \in X} J(w)$$

where

$$X = \{v \in H^1(\Omega) \mid v|_{\Gamma} = 0\}$$

and

$$J(w) = \frac{1}{2} \int_{\Omega} \underbrace{\nabla w \cdot \nabla w}_{w_x^2 + w_y^2} d\Omega - \int_{\Omega} f w d\Omega$$

## Minimization Principle – Proof (sketch)

Let  $w = u + v$ ,  $u \in X$ ,  $v \in X$ , apply Gauss Theorem and boundary conditions to obtain

$$J(\underbrace{u + v}_w) = J(u) + \underbrace{\frac{1}{2} \int_{\Omega} \nabla v \cdot \nabla v d\Omega}_{>0 \text{ unless } v=0}, \quad \forall v \in X$$

$$J(w) > J(u), \quad \forall w \in X, w \neq u$$



$u$  is the minimizer of  $J(w)$

## Weak Formulation

Let

BILINEAR FORM

$$a(w, v) = \int_{\Omega} \nabla w \cdot \nabla v \, d\Omega, \quad \forall w, v \in X$$

and

LINEAR FORM

$$f(v) = \int_{\Omega} f v \, d\Omega, \quad \forall v \in X$$

### Weak Statement

Find  $u \in X$ , such that

$$a(u, v) = f(v), \quad \forall v \in X$$

# Neumann Problem

Find  $u$  such that

STRONG

$$\begin{aligned} -\nabla^2 u &= f && \text{in } \Omega \\ u &= 0 && \text{on } \Gamma^D \\ \frac{\partial u}{\partial n} &= g && \text{on } \Gamma^N \end{aligned}$$

where  $\bar{\Gamma} = \bar{\Gamma}^D \cup \bar{\Gamma}^N$ ,  $\Gamma^D$  non-empty.

Find

$$u = \arg \min_{w \in X} J(w)$$

MINIMIZATION

where

$$X = \{v \in H^1(\Omega) \mid v|_{\Gamma^D} = 0\}$$

$$J(w) = \frac{1}{2} \int_{\Omega} \nabla w \cdot \nabla w d\Omega - \int_{\Omega} f w d\Omega - \int_{\Gamma^N} g w dS$$

## Neumann Problem – Weak Formulation

Let

BILINEAR FORM

$$a(w, v) = \int_{\Omega} \nabla w \cdot \nabla v \, d\Omega, \quad \forall w, v \in X$$

and

LINEAR FORM

$$f(v) = \int_{\Omega} f v \, d\Omega + \int_{\Gamma^N} g v \, dS, \quad \forall v \in X$$

### Weak Statement

Find  $u \in X$ , such that

$$a(u, v) = f(v), \quad \forall v \in X$$



## Summary

- ▶ Minimization/weak formulations are defined by:
  - ▶ a space  $X$ ;
  - ▶ a bilinear form  $a$ ;
  - ▶ a linear form  $\ell$ .
- ▶ Essential (Dirichlet) boundary conditions are imposed by (or reflected in)  $X$
- ▶ Natural (Neumann) boundary conditions are imposed by (or reflected in)  $J$  (or  $a$  and  $f$ ).
- ▶ Points of departure for the finite element method are
  - ▶ the weak form (more generally); or
  - ▶ the minimization statement (if  $a$  is SPD).

## Definitions

### Definition (Linear space)

A space  $Z$  is a linear or vector space if, for any  $\alpha \in \mathbb{R}$ ,  $w, v \in Z$ ,

$$\alpha w + v \in Z$$

Note:  $\mathbb{R}$  denotes the real numbers, and  $\mathbb{N}$  and  $\mathbb{C}$  shall denote the natural and complex numbers, respectively.

### Definition (Inner Product Space)

An inner product space  $Z$  is a linear space equipped with

- ▶ an inner product  $(w, v)_Z$ ,  $\forall w, v \in Z$ , and
- ▶ induced norm  $\|w\|_Z = \sqrt{(w, w)_Z}$ ,  $\forall w \in Z$ .

## Definitions

### Definition (Inner Product)

An inner product  $w, v \in Z \rightarrow (w, v)_Z \in \mathbb{R}$  has to satisfy

► Bilinearity

$$(\alpha w + v, z)_Z = \alpha(w, z)_Z + (v, z)_Z, \quad \forall \alpha \in \mathbb{R}, w, v, z \in Z$$

$$(z, \alpha w + v)_Z = \alpha(z, w)_Z + (z, v)_Z, \quad \forall \alpha \in \mathbb{R}, w, v, z \in Z$$

► Symmetry

$$(w, v)_Z = (v, w)_Z, \quad \forall w, v \in Z$$

► Positivity

$$(w, w)_Z > 0, \quad \forall w \in Z, w \neq 0$$

$$(w, w)_Z = 0, \quad \text{only if } w = 0$$

Cauchy-Schwarz inequality:

$$(w, v)_Z \leq \|w\|_Z \|v\|_Z, \quad \forall w, v \in Z.$$

## Definitions

### Definition (Norm)

A norm is a map  $\|\cdot\| : Z \rightarrow \mathbb{R}$  such that

$$\|w\|_Z > 0 \quad \forall w \in Z, w \neq 0,$$

$$\|\alpha w\|_Z = |\alpha| \|w\|_Z \quad \forall \alpha \in \mathbb{R}, \forall w \in Z,$$

$$\|w + v\|_Z \leq \|w\|_Z + \|v\|_Z \quad \forall w \in Z, \forall v \in Z.$$

Equivalence of norms  $\|\cdot\|_Z$  and  $\|\cdot\|_Y$ : there exist positive constants  $C_1, C_2$  such that

$$C_1 \|v\|_Z \leq \|v\|_Y \leq C_2 \|v\|_Z.$$

## Definitions

### Definition (Cartesian Product Spaces)

Given two inner product spaces  $Z_1$  and  $Z_2$ , we define

$$Z = Z_1 \times Z_2 \equiv \{(w_1, w_2) | w_1 \in Z_1, w_2 \in Z_2\}$$

and given  $w = (w_1, w_2) \in Z$ ,  $v = (v_1, v_2) \in Z$ , we define

$$w + v \equiv (w_1 + v_1, w_2 + v_2).$$

We also equip  $Z$  with the inner product

$$(w, v)_Z = (w_1, v_1)_{Z_1} + (w_2, v_2)_{Z_2}$$

and induced norm

$$\|w\|_Z = \sqrt{(w, w)_Z}.$$

# Linear Forms

## Definition (Linear Functional)

A functional  $g : Z \rightarrow \mathbb{R}$  is a linear functional if, for any  $\alpha \in \mathbb{R}$ ,  $w, v \in Z$

$$g(\alpha w + v) = \alpha g(w) + g(v)$$

A linear form is bounded, or continuous, over  $Z$  if

$$|g(v)| \leq C \|v\|_Z, \quad \forall v \in Z,$$

for some finite real constant  $C$ .

# Dual Spaces

## Definition (Dual Space)

Given  $Z$ , we define the dual space  $Z'$  as the space of all *bounded* linear functionals over  $Z$ . We associate to  $Z'$  the *dual norm*

$$\|g\|_{Z'} = \sup_{v \in Z} \frac{g(v)}{\|v\|_Z}, \quad \forall g \in Z'.$$

## Theorem (Riesz representation)

For any  $g \in Z'$ , there exists a unique  $w_g \in Z$  such that

$$(w_g, v)_Z = g(v), \quad \forall v \in Z.$$

It directly follows that

$$\|g\|_{Z'} = \|w_g\|_Z.$$

## Bilinear Forms

### Definition (Bilinear Form)

A form  $b : Z_1 \times Z_2 \rightarrow \mathbb{R}$  is bilinear if, for any  $\alpha \in \mathbb{R}$ ,

$$b(\alpha w + v, z) = \alpha b(w, z) + b(v, z), \quad \forall w, v \in Z_1, z \in Z_2$$

$$b(z, \alpha w + v) = \alpha b(z, w) + b(z, v), \quad \forall z \in Z_1, w, v \in Z_2$$

The bilinear form  $b : Z \times Z \rightarrow \mathbb{R}$  is

- ▶ symmetric, if

$$b(w, v) = b(v, w), \quad \forall w, v \in Z;$$

- ▶ skew-symmetric, if

$$b(w, v) = -b(v, w), \quad \forall w, v \in Z;$$

- ▶ positive definite, if

$$b(v, v) \geq 0, \quad \forall v \in Z,$$

with equality only for  $v = 0$ .



## Bilinear Forms

The bilinear form  $b : Z \times Z \rightarrow \mathbb{R}$  is

- ▶ positive semidefinite, if

$$b(v, v) \geq 0, \quad \forall v \in Z.$$

We also define, for a general bilinear form  $b : Z \times Z \rightarrow \mathbb{R}$ ,

- ▶ the symmetric part as

$$b_S(w, v) = 1/2 (b(w, v) + b(v, w)), \quad \forall w, v \in Z;$$

- ▶ the skew-symmetric part as

$$b_{SS}(w, v) = 1/2 (b(w, v) - b(v, w)), \quad \forall w, v \in Z.$$

## Bilinear Forms

The bilinear form  $b : Z \times Z \rightarrow \mathbb{R}$  is

- ▶ **coercive** over  $Z$  if

$$\alpha \equiv \inf_{w \in Z} \frac{b(w, w)}{\|w\|_Z^2}$$

is positive;

- ▶ **continuous** over  $Z$  if

$$\gamma \equiv \sup_{w \in Z} \sup_{v \in Z} \frac{b(w, v)}{\|w\|_Z \|v\|_Z}$$

is finite.

## Parametric Linear and Bilinear Forms

We introduce

- ▶  $\mathcal{D} \in \mathbb{R}^P$  : closed bounded parameter domain;
- ▶  $\mu = (\mu_1, \dots, \mu_P) \in \mathcal{D}$  : parameter vector.

We shall say that

- ▶  $g : Z \times \mathcal{D} \rightarrow \mathbb{R}$  is a **parametric linear form** if,  
for all  $\mu \in \mathcal{D}$ ,  $g(\cdot; \mu) : Z \rightarrow \mathbb{R}$  is a linear form;
- ▶  $b : Z \times Z \times \mathcal{D} \rightarrow \mathbb{R}$  is a **parametric bilinear form** if,  
for all  $\mu \in \mathcal{D}$ ,  $b(\cdot, \cdot; \mu) : Z \times Z \rightarrow \mathbb{R}$  is a bilinear form.

Concepts of symmetry, ... directly extend to the parametric case.

## Parametric Linear and Bilinear Forms

The parametric bilinear form  $b : Z \times Z \times \mathcal{D} \rightarrow \mathbb{R}$  is

- ▶ **coercive** over  $Z$  if

$$\alpha(\mu) \equiv \inf_{w \in Z} \frac{b(w, w; \mu)}{\|w\|_Z^2}$$

is positive for all  $\mu \in \mathcal{D}$ ;

- ▶ **continuous** over  $Z$  if

$$\gamma(\mu) \equiv \sup_{w \in Z} \sup_{v \in Z} \frac{b(w, v; \mu)}{\|w\|_Z \|v\|_Z}$$

is finite for all  $\mu \in \mathcal{D}$ .

We also define

$$\begin{aligned} (0 <) \alpha_0 &\equiv \min_{\mu \in \mathcal{D}} \alpha(\mu), \\ \gamma_0 &\equiv \max_{\mu \in \mathcal{D}} \gamma(\mu) (< \infty). \end{aligned}$$

## Coercivity Eigenproblem

We have

$$\alpha(\mu) \equiv \inf_{w \in Z} \frac{b_S(w, w; \mu)}{\|w\|_Z^2};$$

Associated **generalized eigenproblem**:

Given  $\mu \in \mathcal{D}$ , find  $(\chi^{co}, \nu^{co})_i(\mu) \in Z \times \mathbb{R}$ ,  $1 \leq i \leq \dim(Z)$ , such that

$$b_S(\chi_i^{co}(\mu), v; \mu) = \nu_i^{co}(\mu) (\chi_i^{co}(\mu), v)_Z, \quad \forall v \in Z,$$

and

$$\|\chi_i^{co}(\mu)\|_Z = 1.$$

Let  $\nu_1^{co}(\mu) \leq \nu_2^{co}(\mu) \leq \dots \leq \nu_{\dim(Z)}^{co}(\mu)$  and  $b$  coercive, then

$$\alpha(\mu) = \nu_1^{co}(\mu) > 0.$$

## Affine Parameter Dependence

We assume

$$g(v; \mu) = \sum_{q=1}^{Q_g} \Theta_g^q(\mu) g^q(v), \quad \forall v \in Z,$$

where, for  $1 \leq q \leq Q_g$  (finite),

- ▶ parameter-*dependent* functions  $\Theta_g^q : \mathcal{D} \rightarrow \mathbb{R}$ ,
- ▶ parameter-*independent* forms  $g^q : Z \rightarrow \mathbb{R}$ ;

and

$$b(w, v; \mu) = \sum_{q=1}^{Q_b} \Theta_b^q(\mu) b^q(w, v), \quad \forall w, v \in Z,$$

where, for  $1 \leq q \leq Q_b$  (finite),

- ▶ parameter-*dependent* functions  $\Theta_b^q : \mathcal{D} \rightarrow \mathbb{R}$ ,
- ▶ parameter-*independent* forms  $b^q : Z \times Z \rightarrow \mathbb{R}$ .

## Parametric Coercivity

### Definition (Parametric Coercivity)

The coercive bilinear form  $b : Z \times Z \times \mathcal{D} \rightarrow \mathbb{R}$

$$b(w, v; \mu) = \sum_{q=1}^{Q_b} \Theta_b^q(\mu) b^q(w, v), \quad \forall w, v \in Z$$

is **parametrically coercive** if  $c \equiv b_S$  is affine

$$c(w, v; \mu) = \sum_{q=1}^{Q_c} \Theta_c^q(\mu) c^q(w, v), \quad \forall w, v \in Z$$

and satisfies

$$\Theta_c^q(\mu) > 0, \quad \forall \mu \in \mathcal{D}, 1 \leq q \leq Q_c,$$

and

$$c^q(v, v) \geq 0, \quad \forall v \in Z, 1 \leq q \leq Q_c.$$

## Scalar and Vector Field

We consider (real)

- ▶ scalar-valued field variables (e.g., temperature, pressure)

$$w : \Omega \rightarrow \mathbb{R}^{d_v=1}$$

- ▶ vector-valued field variables (e.g., displacement, velocity)

$$w : \Omega \rightarrow \mathbb{R}^{d_v},$$

where  $w(x) = (w_1(x), \dots, w_{d_v}(x))$ ;

and

- ▶  $\Omega \in \mathbb{R}^d$ ,  $d = 1, 2$ , or  $3$  is an open bounded; domain
- ▶  $x = (x_1, \dots, x_d) \in \Omega$ ;
- ▶  $\Omega$  has Lipschitz continuous boundary  $\partial\Omega$ ; and
- ▶ we define the canonical basis vectors as  $e_i$ ,  $1 \leq i \leq d$ .



## Multi-index Derivatives

Given a scalar (or one component of a vector)

- ▶ field  $w : \Omega \rightarrow \mathbb{R}$  SPATIAL DERIVATIVE

$$(D^\sigma w)(x) = \frac{\partial^\sigma w}{\partial x_1^{\sigma_1} \dots \partial x_d^{\sigma_d}}$$

- ▶ **parametric** field  $w : \Omega \times \mathcal{D} \rightarrow \mathbb{R}$  SENSITIVITY DERIVATIVE

$$(D_\sigma w)(x) = \frac{\partial^\sigma w}{\partial \mu_1^{\sigma_1} \dots \partial \mu_d^{\sigma_d}}$$

where

- ▶  $\sigma = (\sigma_1, \dots, \sigma_d)$ ,  $\sigma_i$ ,  $1 \leq i \leq d$ , non-negative integers;
- ▶  $|\sigma| = \sum_{j=1}^d \sigma_j$  is the order of the derivative; and
- ▶  $I^{d,n}$  is set of all index vectors  $\sigma \in \mathbb{N}_0^d$  such that  $|\sigma| \leq n$ .

## Function Spaces

### Definition (Spaces of Continuous Functions)

Let  $m \in \mathbb{N}_0$ , the space  $C^m(\Omega)$  is defined as

$$C^m(\Omega) \equiv \{w \mid D^\sigma w \in C^0(\Omega), \forall \sigma \in I^{d,m}\},$$

and  $C^0(\Omega)$  is the space of continuous functions over  $\Omega \in \mathbb{R}^d$ .

- ▶ We denote by  $C^\infty(\Omega)$  the space of functions  $w$  for which  $D^\sigma$  exists and is continuous for any order  $|\sigma|$ .

# Lebesgue Spaces

## Definition (Lebesgue Spaces)

We define, for  $1 \leq p < \infty$ , the Lebesgue space  $L^p(\Omega)$  as

$$L^p(\Omega) \equiv \{w \text{ measurable} \mid \|w\|_{L^p(\Omega)} < \infty\}$$

where

$$\|w\|_{L^p(\Omega)} \equiv \left( \int_{\Omega} |w|^p dx \right)^{\frac{1}{p}}, \quad 1 \leq p < \infty,$$

$$\|w\|_{L^\infty(\Omega)} \equiv \operatorname{ess\,sup}_{x \in \Omega} |w(x)|, \quad p = \infty.$$

# Hilbert Space

## Definition (Hilbert space)

Let  $m \in \mathbb{N}_0$ , the space  $H^m(\Omega)$  is then defined as

$$H^m(\Omega) \equiv \{w \mid D^\sigma w \in L^2(\Omega), \forall \sigma \in I^{d,m}\},$$

with associated inner product

$$(w, v)_{H^m(\Omega)} \equiv \sum_{\sigma \in I^{d,m}} \int_{\Omega} D^\sigma w D^\sigma v \, dx,$$

and induced norm

$$\|w\|_{H^m(\Omega)} \equiv \sqrt{(w, w)_{H^m(\Omega)}}.$$

## Special (Most Important) Cases

Since we only consider **second-order** PDEs, we require mostly

- ▶  $L^2(\Omega) = H^0(\Omega)$ : Lebesgue Space  $p = 2$

$$L^2(\Omega) \equiv \{w \text{ measurable} \mid \int_{\Omega} w^2 < \infty\}$$

with inner product and induced norm

$$(w, v)_{L^2(\Omega)} \equiv \int_{\Omega} w v, \quad \forall w, v \in L^2(\Omega)$$

$$\|w\|_{L^2(\Omega)} \equiv \sqrt{(w, w)_{L^2(\Omega)}}, \quad \forall w \in L^2(\Omega),$$

⇒ Space of all functions  $w : \Omega \rightarrow \mathbb{R}$  square-integrable over  $\Omega$ .

## Special (Most Important) Cases

Since we only consider **second-order** PDEs, we require mostly

►  $H^1(\Omega)$

$$H^1(\Omega) \equiv \{w \in L^2(\Omega) \mid \frac{\partial w}{\partial x_i} \in L^2(\Omega), 1 \leq i \leq d\}$$

with inner product and induced norm

$$(w, v)_{H^1(\Omega)} \equiv \int_{\Omega} \nabla w \cdot \nabla v + wv, \quad \forall w, v \in H^1(\Omega),$$

$$\|w\|_{H^1(\Omega)} \equiv \sqrt{(w, w)_{H^1(\Omega)}}, \quad \forall w \in H^1(\Omega),$$

and seminorm

$$|w|_{H^1(\Omega)} \equiv \int_{\Omega} \nabla w \cdot \nabla w, \quad \forall w \in H^1(\Omega).$$

## Special (Most Important) Cases

Since we only consider **second-order** PDEs, we require mostly

- ▶ and the space

$$H_0^1(\Omega) \equiv \{v \in H^1(\Omega) \mid v|_{\partial\Omega} = 0\}$$

where  $v = 0$  on the boundary  $\partial\Omega$ .

Note that, for any  $v \in H_0^1(\Omega)$ , we have

$$C_{PF} \|v\|_{H^1(\Omega)} \leq |v|_{H^1(\Omega)} \leq \|v\|_{H^1(\Omega)},$$

and thus

$$|v|_{H^1(\Omega)} = 0 \text{ implies } v = 0.$$

$\Rightarrow |v|_{H^1(\Omega)}$  constitutes a norm for  $v \in H_0^1(\Omega)$ .

# Projection

## Definition (Projection)

Given Hilbert Spaces  $Y$  and  $Z \subset Y$ , the projection,  $\Pi : Y \rightarrow Z$ , of  $y \in Y$  onto  $Z$  is defined as

$$(\Pi y, v)_Y = (y, v)_Y, \quad \forall v \in Z$$

### Properties

- ▶ Orthogonality:  $(y - \Pi y, v)_Y = 0, \quad \forall v \in Z$
- ▶ Idempotence:  $\Pi(\Pi y) = \Pi y$
- ▶ Best Approximation

$$\|y - \Pi y\|_Y^2 = \inf_{v \in Z} \|y - v\|_Y^2$$

- ▶ Given an orthonormal basis  $\{\phi_i\}_{i=1}^N$ ,  $N = \dim(Z)$ , then

$$\Pi y = \sum_{i=1}^N (\phi_i, y)_Y \phi_i, \quad \forall y \in Y$$



## General Preliminaries

- The Poisson Problem
- Inner Product Spaces
- Linear and Bilinear Forms
- Classes of Functions
  - Scalar and Vector Fields
  - Function Spaces

## Linear Elliptic Problems I

- Problem Statement
- Truth Approximation

## Reduced Basis Approximation

- Objective
- RB Approximation Space
- Galerkin Projection
- Offline-Online Computational Procedure

## Problem Statement

Given  $\mu \in \mathcal{D} \subset \mathbb{R}^P$ , evaluate

*P*-vector INPUT

$$s^e(\mu) = \ell(u^e(\mu); \mu)$$

OUTPUT

where  $u^e(x; \mu) \in X^e(\Omega)$  satisfies

FIELD ( $x; \mu$ )

$$a(u^e(\mu), v; \mu) = f(v; \mu), \quad \forall v \in X^e(\Omega).$$

$\mu$ PDE

- ▶ Here <sup>e</sup> refers to “exact,” i.e., the exact (or analytic) solution to the problem for the prescribed mathematical model.
- ▶ To begin, we consider elliptic problems which are linear, coercive, affine, and compliant — explanations to follow . . .

## Definitions

- $\mu$ : input parameter -  $\mu = (\mu_1, \mu_2, \dots, \mu_P)$ ;  $P$ -tuple
- $\mathcal{D}$ : parameter domain in  $\mathbb{R}^P$ ;
- $\Omega$ : spatial domain in  $\mathbb{R}^d$ ;
- $s^e$ : output;
- $\ell$ : output functional;
- $u^e$ : field variable;
- $X^e$ : function space  $(H_0^1(\Omega))^\nu \subset X^e \subset (H^1(\Omega))^\nu$  †,  
 with inner product  $(w, v)_{X^e}, \forall w, v \in X^e$ ,  
 and induced norm  $\|w\|_{X^e} = \sqrt{(w, w)_{X^e}}, \forall w \in X^e$ .

---

† For simplicity we first assume  $\nu = 1$ .

## Reference Geometry

Note  $\Omega$  is **parameter-independent**:

- ▶ the reduced basis requires a common spatial configuration, i.e., a reference domain  $\Omega_{\text{ref}}$
- ▶ Introduce a piecewise affine mapping  $\mathcal{T}(\cdot; \mu) : \Omega \rightarrow \Omega_o(\mu)$

$$\begin{array}{ccc}
 a_o(w_o, v_o; \mu) \text{ over } \Omega_o(\mu) & & \\
 \Downarrow & & \\
 \mathcal{T}(\cdot; \mu)^{-1} : \Omega_o(\mu) \rightarrow \Omega_{\text{ref}} \equiv \Omega & & (\Omega_{\text{ref}} = \Omega_o(\mu_{\text{ref}})) \\
 \Downarrow & & \\
 a(w, v; \mu) \text{ over } \Omega & & 
 \end{array}$$

where  $a(w, v; \mu) = a_o(w_o \circ \mathcal{T}_\mu, v_o \circ \mathcal{T}_\mu; \mu)$

We will discuss this issue in detail later on.

## Hypotheses – Continuity, Stability, Compliance

$$\left. \begin{array}{l}
 a(\cdot, \cdot; \mu) : \text{ bilinear,} \\
 \text{ symmetric,} \\
 X^e\text{-continuous,} \\
 X^e\text{-coercive form, } \forall \mu \in \mathcal{D}; \\
 \\
 f(\cdot; \mu), \ell(\cdot; \mu) : \text{ linear,} \\
 X^e\text{-bounded, } \forall \mu \in \mathcal{D}
 \end{array} \right\} \mu\text{PDE}$$

We first consider the **compliant** case:

- ▶  $\ell(\cdot; \mu) = f(\cdot; \mu), \forall \mu \in \mathcal{D}$
- ▶  $a$  symmetric

## Hypotheses – Affine parameter dependence

Require

also  $f(v; \mu)$ ,  $\ell(v; \mu)$

$$a(w, v; \mu) = \sum_{q=1}^{Q_a} \Theta_a^q(\mu) a^q(w, v),$$

where for  $q = 1, \dots, Q_a$

$$\begin{aligned} \Theta_a^q &: \mathcal{D} \rightarrow \mathbb{R}, & \mu\text{-dependent functions;} \\ a^q &: X^e \times X^e \rightarrow \mathbb{R}, & \mu\text{-independent forms.} \end{aligned}$$

This assumption

- ▶ is broadly applicable to many instances of property and geometry parametric variations, but
- ▶ can also be relaxed (see lectures on "nonaffine problems").

## Inner Products and Norms

- ▶ Energy inner product and induced norm (**parameter-dependent**)

$$((w, v))_{\mu} = a(w, v; \mu), \quad \forall w, v \in X^e$$

$$|||w|||_{\mu} = \sqrt{a(w, w; \mu)}, \quad \forall w \in X^e$$

- ▶  $X^e$ -inner product and induced norm (**parameter-independent**)

$$(w, v)_{X^e} \equiv ((w, v))_{\bar{\mu}} (= a(w, v; \bar{\mu})), \quad \forall w, v \in X^e$$

$$\|w\|_{X^e} \equiv |||w|||_{\bar{\mu}} (= \sqrt{a(w, w; \bar{\mu})}), \quad \forall w \in X^e$$

## Coercivity and Continuity constants

Recall

- ▶ Coercivity constant

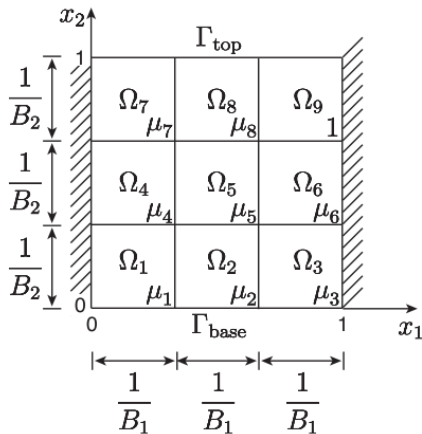
$$(0 <) \alpha^e(\mu) \equiv \inf_{w \in X^e} \frac{a(w, w; \mu)}{\|w\|_{X^e}^2};$$

- ▶ Continuity constant

$$\gamma^e(\mu) \equiv \sup_{w \in X^e} \sup_{v \in X^e} \frac{a(w, v; \mu)}{\|w\|_{X^e} \|v\|_{X^e}} (< \infty).$$



## Example: ThermalBlock



$$\bar{\Omega} = \cup_{i=1}^{B_1 B_2} \bar{\Omega}_i$$

## Example: ThermalBlock

### Problem Statement

Given  $\mu = (\mu_1, \dots, \mu_P) \subset \mathcal{D} \equiv [\mu^{\min}, \mu^{\max}]^P$ , evaluate

$$s^e(\mu) = f(u^e(\mu)) \quad \ell = f$$

where  $u^e(\mu) \in X^e \equiv \{v \in H^1(\Omega) \mid v|_{\Gamma_{\text{top}}} = 0\}$  satisfies

$$a(u^e(\mu), v; \mu) = f(v), \quad \forall v \in X^e.$$

Here,  $P = B_1 B_2 - 1$ , and we assume that

$\mu^{\min} = 1/\sqrt{\mu_r}$ ,  $\mu^{\max} = \sqrt{\mu_r}$ , for  $1 < \mu_r < \infty$ ,  
 such that  $\mu^{\max}/\mu^{\min} = \mu_r$ .

## Example: ThermalBlock

Here,  $\forall v \in X^e$ ,

compliant  $\ell = f$

$$f(v) \equiv \int_{\Gamma_{\text{base}}} v,$$

and,  $\forall w, v \in X^e$ ,

$$a(w, v; \mu) = \sum_{i=1}^P \mu_i \int_{\Omega_i} \nabla w \cdot \nabla v + \int_{\Omega_{P+1}} \nabla w \cdot \nabla v,$$

where  $\bar{\Omega} = \cup_{i=1}^{P+1} \bar{\Omega}_i$ .

## Example: ThermalBlock

Inner product,  $w, v \in X^e$ ,

$$(w, v)_{X^e} = \sum_{i=1}^P \bar{\mu}_i \int_{\Omega_i} \nabla w \cdot \nabla v + \int_{\Omega_{P+1}} \nabla w \cdot \nabla v,$$

where  $\bar{\mu}_i$  is the **reference parameter**.

The bilinear form  $a$  is

- ▶ **symmetric**,
- ▶ **(parametrically) coercive**,  
 $0 < \frac{1}{\sqrt{\mu_r}} \leq \text{Min}(\mu_1/\bar{\mu}_1, \dots, \mu_P/\bar{\mu}_P, 1) \leq \alpha^e(\mu),$
- ▶ and **continuous**,  
 $\gamma^e(\mu) < \text{Max}(\mu_1/\bar{\mu}_1, \dots, \mu_P/\bar{\mu}_P, 1) \leq \sqrt{\mu_r} < \infty.$

The linear form  $f$  is **bounded**.

## Example: ThermalBlock

We obtain

$$a(w, v; \mu) = \sum_{q=1}^{Q_a=P+1} \Theta^q(\mu) a^q(w, v):$$

for

$$\Theta^1(\mu) = \mu_1, \quad a^1(w, v) = \int_{\Omega_1} \nabla w \cdot \nabla v ;$$

$$\vdots$$

$$\Theta^P(\mu) = \mu_P, \quad a^P(w, v) = \int_{\Omega_P} \nabla w \cdot \nabla v ;$$

$$\Theta^{P+1}(\mu) = 1, \quad a^{P+1}(w, v) = \int_{\Omega_{P+1}} \nabla w \cdot \nabla v .$$

$\Rightarrow$  affine assumption is honored

## Approximation Space

We replace the “exact” statement by a “truth” approximation, i.e., we introduce

- ▶ an approximation space  $X^{\mathcal{N}} \subset X^e$  of dimension  $\dim(X^{\mathcal{N}}) = \mathcal{N}$ , and
- ▶ an associated set of basis functions  $\varphi_k^{\mathcal{N}} \in X^{\mathcal{N}}$ ,  $1 \leq k \leq \mathcal{N}$ .

We also equip  $X^{\mathcal{N}}$  with inner product and induced norm

$$(w, v)_{X^{\mathcal{N}}} \equiv (w, v)_{X^e} \equiv a(w, v; \bar{\mu}), \quad \forall w, v \in X^{\mathcal{N}}$$

$$\|w\|_{X^{\mathcal{N}}} \equiv \|w\|_{X^e} \equiv \sqrt{a(w, w; \bar{\mu})}, \quad \forall w \in X^{\mathcal{N}}$$

## Approximation Space

- ▶ Coercivity constant of  $a$  over  $X^{\mathcal{N}}$

$$\alpha^{\mathcal{N}}(\mu) \equiv \inf_{w \in X^{\mathcal{N}}} \frac{a(w, w; \mu)}{\|w\|_{X^{\mathcal{N}}}^2};$$

- ▶ Continuity constant of  $a$  over  $X^{\mathcal{N}}$

$$\gamma^{\mathcal{N}}(\mu) \equiv \sup_{w \in X^{\mathcal{N}}} \sup_{v \in X^{\mathcal{N}}} \frac{a(w, v; \mu)}{\|w\|_{X^{\mathcal{N}}} \|v\|_{X^{\mathcal{N}}}}.$$

**Note:** since  $X^{\mathcal{N}} \subset X^e$ ,

$$\alpha^{\mathcal{N}}(\mu) \geq \alpha^e(\mu) (> 0), \quad \forall \mu \in \mathcal{D},$$

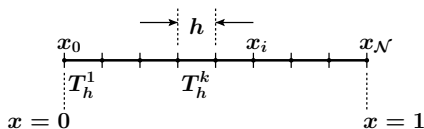
$$\gamma^{\mathcal{N}}(\mu) \leq \gamma^e(\mu) (< \infty), \quad \forall \mu \in \mathcal{D}.$$

## Approximation Space – Finite Element

We consider linear ( $\mathbf{IP}_1$ ) or quadratic ( $\mathbf{IP}_2$ ) Finite Element Spaces

**Example:** Consider the domain  $\Omega \equiv ]0, 1[ \subset \mathbb{R}^{d=1}$

- ▶ Triangulation  $\mathcal{T}_h$  with segments  $T_h^j$ ,  $1 \leq j \leq \mathcal{N}$



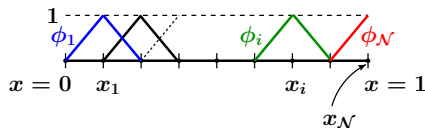
Properties

- ▶  $T_h^j$  are open (closure  $\bar{T}_h^j$ )
- ▶  $T_h^j \cap T_h^i = \emptyset$  for  $j \neq i$
- ▶  $\bar{\Omega} = \bigcup_{T_h \in \mathcal{T}_h} \bar{T}_h$



## Approximation Space – Finite Element

- ▶ Nodal basis functions  $\varphi_i(x)$ ,  $1 \leq i \leq \mathcal{N}$



### Properties

- ▶ Compact support in  $\Omega$  (i.e.,  $\varphi_i$  nonzero only on  $\overline{T}_h^i \cup \overline{T}_h^{i+1}$ )
- ▶ For all  $v \in X^{\mathcal{N}}$ ,  $\exists$  unique  $v_n \in \mathbb{R}$ , such that
 
$$v(x) = \sum_{n=1}^{\mathcal{N}} v_n \varphi_n(x)$$
- ▶ "Truth" finite element approximation space
 
$$X^{\mathcal{N}} = \{v \in X^e \mid v|_{T_h^j} \in \mathbb{P}_1(T_h^j), 1 \leq j \leq \mathcal{N}\};$$
 (Space of functions which are linear over each element)

## "Truth" Problem Statement

Given  $\mu \in \mathcal{D} \subset \mathbb{R}^P$ , evaluate

Galerkin projection

$$s^{\mathcal{N}}(\mu) = f(u^{\mathcal{N}}(\mu); \mu)$$

where  $u^{\mathcal{N}}(x; \mu) \in X^{\mathcal{N}}(\Omega)$  satisfies

$$a(u^{\mathcal{N}}(\mu), v; \mu) = f(v; \mu), \quad \forall v \in X^{\mathcal{N}}(\Omega).$$

**Assumption:**

No variational crimes in Galerkin approximation, i.e.,

- ▶ truth approximation represents exact geometry,
- ▶ all quadratures are exact.

## "Truth" Problem Statement

Galerkin projection convergent:

- ▶  $u^{\mathcal{N}}(\mu) \rightarrow u^e(\mu)$  as  $\mathcal{N} \rightarrow \infty$ ; and ( $f$  is bounded)
- ▶  $s^{\mathcal{N}}(\mu) \rightarrow s^e(\mu)$  as  $\mathcal{N} \rightarrow \infty$ .

We choose  $\mathcal{N}$  such that

$\mathcal{N}$  *conservative*

$$\varepsilon^{\mathcal{N}} = \max_{\mu \in \mathcal{D}} \|u^e(\mu) - u^{\mathcal{N}}(\mu)\|_{X^e} \ll \epsilon_{\text{tol, min}},$$

and hence we require that our RB formulation be

- (a) numerically stable as  $\mathcal{N} \rightarrow \infty$ ; and
- (b) computationally efficient as  $\mathcal{N} \rightarrow \infty$ .

## Truth Approximation – Role

We shall

- (i) *build* our reduced basis approximation upon “truth” solutions  $u^{\mathcal{N}}(\mu) \in X^{\mathcal{N}}$ ;
- (ii) *measure* the error in the reduced basis approximation relative to the “truth” solution  $u^{\mathcal{N}}(\mu) \in X^{\mathcal{N}}$  (and  $s^{\mathcal{N}}(\mu)$ );

$\Rightarrow u^{\mathcal{N}}(\mu)$  is a *calculable surrogate* for  $u^e(\mu)$ .

$$\|u^e(\mu) - u_N(\mu)\|_{X^e} \leq \underbrace{\|u^e(\mu) - u^{\mathcal{N}}(\mu)\|_{X^e}}_{\leq \varepsilon_{\text{exact}}} + \underbrace{\|u^{\mathcal{N}}(\mu) - u_N(\mu)\|_{X^e}}_{\leq \varepsilon_{\text{tol,min}}}$$

with  $\varepsilon_{\text{exact}} \ll \varepsilon_{\text{tol,min}}$ .

## Algebraic Equations

Expand 
$$\mathbf{u}^{\mathcal{N}}(x; \mu) = \sum_{j=1}^{\mathcal{N}} \mathbf{u}_j^{\mathcal{N}}(\mu) \varphi_j^{\mathcal{N}}(x)$$

Given  $\mu \in \mathcal{D}$ , evaluate

$$\mathbf{s}^{\mathcal{N}}(\mu) = (\underline{\mathbf{F}}^{\mathcal{N}}(\mu))^T \underline{\mathbf{u}}^{\mathcal{N}}(\mu)$$

where  $\underline{\mathbf{u}}^{\mathcal{N}}(\mu) \equiv [\mathbf{u}_1^{\mathcal{N}}(\mu) \ \mathbf{u}_2^{\mathcal{N}}(\mu) \ \dots \ \mathbf{u}_{\mathcal{N}}^{\mathcal{N}}(\mu)]^T \in \mathbb{R}^{\mathcal{N}}$  satisfies

$$\underline{\mathbf{A}}^{\mathcal{N}}(\mu) \underline{\mathbf{u}}^{\mathcal{N}}(\mu) = \underline{\mathbf{F}}^{\mathcal{N}}(\mu).$$

- ▶ Stiffness matrix  $\underline{\mathbf{A}}^{\mathcal{N}}(\mu) \in \mathbb{R}^{\mathcal{N} \times \mathcal{N}}$

$$A_{ij}^{\mathcal{N}}(\mu) = a(\varphi_j^{\mathcal{N}}, \varphi_i^{\mathcal{N}}; \mu), \quad 1 \leq i, j \leq \mathcal{N}$$

- ▶ Load/source vector  $\underline{\mathbf{F}}^{\mathcal{N}}(\mu) \in \mathbb{R}^{\mathcal{N}}$

$$F_i^{\mathcal{N}}(\mu) = f(\varphi_i^{\mathcal{N}}; \mu), \quad 1 \leq i \leq \mathcal{N}$$

# Algebraic Equations

## Affine Assumption

- ▶ Stiffness matrix:  $\underline{\mathbf{A}}^{\mathcal{N}}(\mu) = \sum_{q=1}^{Q_a} \Theta_a^q(\mu) \underline{\mathbf{A}}^{\mathcal{N}q}$ , where  
 $\underline{\mathbf{A}}^{\mathcal{N}q} \in \mathbb{R}^{\mathcal{N} \times \mathcal{N}}$ ,  $1 \leq q \leq Q_a$ , and  
 $\underline{\mathbf{A}}_{ij}^{\mathcal{N}q} = a^q(\varphi_j^{\mathcal{N}}, \varphi_i^{\mathcal{N}})$ ,  $1 \leq i, j \leq \mathcal{N}$ ,  $1 \leq q \leq Q_a$ .
- ▶ Load/source vector:  $\underline{\mathbf{F}}^{\mathcal{N}}(\mu) = \sum_{q=1}^{Q_f} \Theta_f^q(\mu) \underline{\mathbf{F}}^{\mathcal{N}q}$ , where  
 $\underline{\mathbf{F}}^{\mathcal{N}q} \in \mathbb{R}^{\mathcal{N}}$ ,  $1 \leq q \leq Q_f$ , and  
 $\underline{\mathbf{F}}_i^{\mathcal{N}q} = f^q(\varphi_i^{\mathcal{N}})$ ,  $1 \leq i \leq \mathcal{N}$ ,  $1 \leq q \leq Q_f$ .

Note: the  $\underline{\mathbf{A}}^{\mathcal{N}q}$  and  $\underline{\mathbf{F}}^{\mathcal{N}q}$  are **parameter-independent**.

# Algebraic Equations

We also introduce  $\underline{\mathbb{X}}^{\mathcal{N}} \in \mathbb{R}^{\mathcal{N} \times \mathcal{N}}$ , such that

$$\underline{\mathbb{X}}_{ij}^{\mathcal{N}} = (\varphi_j^{\mathcal{N}}, \varphi_i^{\mathcal{N}})_{\mathbf{X}^{\mathcal{N}}}, \quad 1 \leq i, j \leq \mathcal{N}$$

Then, given  $w, v \in \mathbf{X}^{\mathcal{N}}$ , we expand

$$w = \sum_{j=1}^{\mathcal{N}} w_j \varphi_j^{\mathcal{N}} \quad \text{and} \quad v = \sum_{j=1}^{\mathcal{N}} v_j \varphi_j^{\mathcal{N}},$$

and thus

$$(w, v)_{\mathbf{X}^{\mathcal{N}}} = \sum_{j=1}^{\mathcal{N}} \sum_{i=1}^{\mathcal{N}} w_j v_i (\varphi_j^{\mathcal{N}}, \varphi_i^{\mathcal{N}})_{\mathbf{X}^{\mathcal{N}}} = \underline{w}^T \underline{\mathbb{X}}^{\mathcal{N}} \underline{v},$$

where

$$\underline{w} \equiv [w_1 \ w_2 \ \dots \ w_{\mathcal{N}}]^T \in \mathbb{R}^{\mathcal{N}} \quad \text{and} \quad \underline{v} \equiv [v_1 \ v_2 \ \dots \ v_{\mathcal{N}}]^T \in \mathbb{R}^{\mathcal{N}}.$$

## Nomenclature

For the rest of this course we will use the following notation

$()^e$  exact problem statement

$\mathbf{X}^e, \mathbf{u}^e, \mathbf{s}^e, \dots$

$() = ()^{\mathcal{N}}$  truth approximation

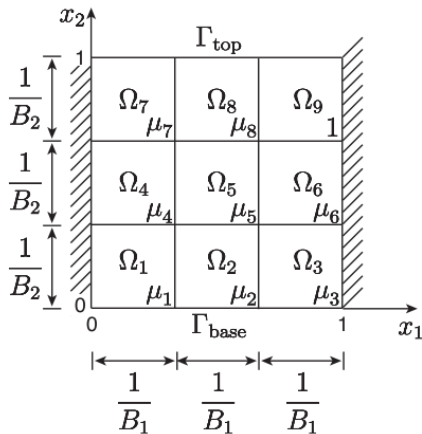
e.g.  $\mathbf{X} = \mathbf{X}^{\mathcal{N}}, \mathbf{u} = \mathbf{u}^{\mathcal{N}}, \mathbf{s} = \mathbf{s}^{\mathcal{N}}, \dots$

$()_N = ()_N^{\mathcal{N}}$  reduced basis approximation

e.g.  $\mathbf{X}_N = \mathbf{X}_N^{\mathcal{N}}, \mathbf{u}_N = \mathbf{u}_N^{\mathcal{N}}, \mathbf{s}_N = \mathbf{s}_N^{\mathcal{N}}, \dots$



## Example: thermal block



$$\bar{\Omega} = \cup_{i=1}^{B_1 B_2} \bar{\Omega}_i$$

## Example: thermal block

### “Truth” Problem Statement

Given  $\mu = (\mu_1, \dots, \mu_P) \in \mathcal{D} \equiv [\mu^{\min}, \mu^{\max}]^P$ , evaluate

$$s^{\mathcal{N}}(\mu) = f(u^{\mathcal{N}}(\mu)) \quad \ell = f$$

where  $u^{\mathcal{N}}(\mu) \in X^{\mathcal{N}}$  satisfies

$$a(u^{\mathcal{N}}(\mu), v; \mu) = f(v), \quad \forall v \in X^{\mathcal{N}}.$$

Here,  $P = B_1 B_2 - 1$  and we require  $0 < \mu^{\min} < \mu^{\max} < \infty$ .

## Example: thermal block

### “Truth” Problem Statement

Given  $\mu = (\mu_1, \dots, \mu_P) \subset \mathcal{D} \equiv [\mu^{\min}, \mu^{\max}]^P$ , evaluate

$$s(\mu) = f(u(\mu)) \qquad \ell = f$$

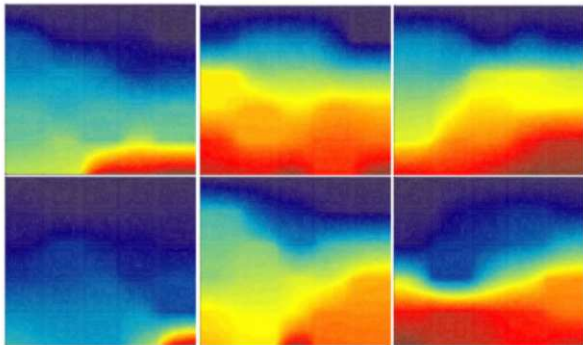
where  $u(\mu) \in X$  satisfies

$$a(u(\mu), v; \mu) = f(v), \quad \forall v \in X.$$

Here,  $P = B_1 B_2 - 1$  and we require  $0 < \mu^{\min} < \mu^{\max} < \infty$ .

## Example: thermal block

Representative Solution ( $P = 24$ )



Source: S. Sen, NHT Part B

## General Preliminaries

- The Poisson Problem
- Inner Product Spaces
- Linear and Bilinear Forms
- Classes of Functions
  - Scalar and Vector Fields
  - Function Spaces

## Linear Elliptic Problems I

- Problem Statement
- Truth Approximation

## Reduced Basis Approximation

- Objective
- RB Approximation Space
- Galerkin Projection
- Offline-Online Computational Procedure

## Reduced Basis (RB) Objective

For **any**  $\varepsilon_{\text{des}} > 0$ , evaluate ACCURACY

$$\mu \in \mathcal{D} \rightarrow s_N(\mu) (\approx s^{\mathcal{N}}(\mu)) \text{ and } \Delta_N^s(\mu)$$

that **provably** achieves desired accuracy RELIABILITY

$$|s^{\mathcal{N}}(\mu) - s_N(\mu)| \leq \Delta_N^s(\mu) \leq \varepsilon_{\text{des}}$$

but at (very low) marginal cost  $\partial t_{\text{comp}}$  EFFICIENCY

**independent** of  $\mathcal{N}$  as  $\mathcal{N} \rightarrow \infty$ .

Here,  $\partial t_{\text{comp}}$  is the time to perform

**one additional certified** evaluation  $\mu \rightarrow (s_N(\mu), \Delta_N^s(\mu))$ .

## RB Objective – Rapid Convergence

Find a reduced basis approximation

$$s_N(\mu) \in \mathbb{R} \text{ and } \mathbf{u}_N(\mu) \in \mathbf{X}_N \subset \mathbf{X}^{\mathcal{N}}$$

for all  $\mu \in \mathcal{D}$ , such that

$$s_N(\mu) \rightarrow s^{\mathcal{N}}(\mu) \text{ and } \mathbf{u}_N(\mu) \rightarrow \mathbf{u}^{\mathcal{N}}(\mu)$$

rapidly as  $N = \dim(\mathbf{X}_N) \rightarrow \infty (= 10 - 200)$ .

Note that the convergence rate should be **independent** of  $\mathcal{N}$ .

## RB Objective – Rigor & Certainty

*A posteriori* error bounds  $\Delta_N(\mu)$  and  $\Delta_N^s(\mu)$ :

$$1 \text{ (rigor)} \leq \frac{\Delta_N(\mu)}{\|u^{\mathcal{N}}(\mu) - u_N(\mu)\|_X} \leq C \text{ (sharpness)}$$

and

$$1 \text{ (rigor)} \leq \frac{\Delta_N^s(\mu)}{|s^{\mathcal{N}}(\mu) - s_N(\mu)|} \leq C \text{ (sharpness)}$$

for all  $N \in \mathbb{N} \equiv \{1, \dots, N_{\max}\}$  and all  $\mu \in \mathcal{D}$ .



## RB Objective – Computational Efficiency

Offline-Online computational strategies:

Offline: expensive preprocessing

$$t_{\text{comp}}^{\text{Offline}} \gg \text{cost}\{\mu \rightarrow s^{\mathcal{N}}(\mu)\}$$

BUT

Online: very rapid certified RB input-output evaluation

$$\partial t_{\text{comp}} \equiv \text{marginal cost}\{\mu \rightarrow s_N(\mu), \Delta_N^s(\mu)\}$$

depends only on  $Q$  and  $N$  – not on  $\mathcal{N}$ .

$\Rightarrow$  we may thus choose  $\mathcal{N}$  very conservatively.

## RB Objective – Relevance

Real-Time Context (parameter estimation, ...):

$$\begin{array}{ccc} \mu & \rightarrow & s_N(\mu), \Delta_N^s(\mu). \\ t_0 \text{ ("need")} & & t_0 + \partial t_{\text{comp}} \text{ ("response")} \end{array}$$

Many-Query Context (design, ...):

$$\begin{array}{ccc} \mu_j & \rightarrow & (s_N(\mu_j), \Delta_N^s(\mu_j)), \quad j = 1, \dots, J. \\ t_0 & & t_0 + \partial t_{\text{comp}} J \text{ as } J \rightarrow \infty \end{array}$$

$\Rightarrow$  Low marginal (real-time) and/or low average (many-query) cost.

## Parametric Manifold $\mathcal{M}^{\mathcal{N}}$

We assume

- ▶ the form  $a$  is continuous and coercive (or inf-sup stable); and
- ▶ affine  $\mu$ -dependence; and
- ▶ the  $\Theta^q(\mu)$ ,  $1 \leq q \leq Q$ , are smooth (i.e.,  $\Theta^q \in C^\infty(\mathcal{D})$ );

then

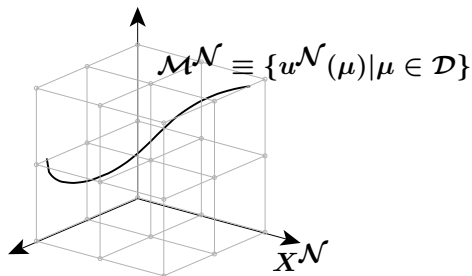
$$\mathcal{M}^{\mathcal{N}} \equiv \{u^{\mathcal{N}}(\mu) \mid \forall \mu \in \mathcal{D}\}$$

is a **smooth**  $P$ -dimensional manifold in  $X^{\mathcal{N}}$ , since

$$\|D_\sigma y^{\mathcal{N}}(\mu)\|_X \leq C_\sigma, \quad \forall \mu \in \mathcal{D}, \text{ for any order } |\sigma| \in \mathbb{N}_{+0}.$$

# Parametric Manifold $\mathcal{M}^{\mathcal{N}}$

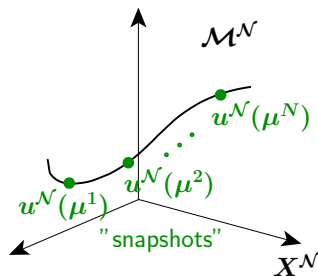
To approximate  $u^{\mathcal{N}}(\mu)$ , and hence  $s^{\mathcal{N}}(\mu)$ ,  
 we need **not** represent **every possible function** in  $X^{\mathcal{N}}$ .



Early work: [ASB], [NP], [FR], [Po], [Pe], [G], [IR], ...

# Parametric Manifold $\mathcal{M}^{\mathcal{N}}$

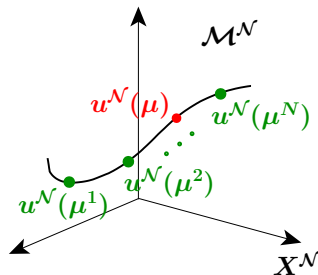
To approximate  $u^{\mathcal{N}}(\mu)$ , and hence  $s^{\mathcal{N}}(\mu)$ ,  
we need **not** represent **every possible function** in  $X^{\mathcal{N}}$ .



LOCALIZATION

# Parametric Manifold $\mathcal{M}^{\mathcal{N}}$

To approximate  $u^{\mathcal{N}}(\mu)$ , and hence  $s^{\mathcal{N}}(\mu)$ ,  
we need **not** represent **every possible function** in  $X^{\mathcal{N}}$ .



SMOOTHNESS

## Spaces & Bases

We define the RB approximation space

$$X_N = \text{span}\{\xi^n, 1 \leq n \leq N\}, \quad 1 \leq N \leq N_{\max},$$

with linearly independent basis functions

$$\xi^n \in X, \quad 1 \leq n \leq N_{\max}.$$

We thus obtain

$$X_N \subset X, \quad \dim(X_N) = N, \quad 1 \leq N \leq N_{\max},$$

and

“nested” (hierarchical) spaces

$$X_1 \subset X_2 \subset \dots \subset X_{N_{\max}-1} \subset X_{N_{\max}} (\subset X).$$

We denote non-hierarchical RB spaces as  $X_N^{\text{nh}}$ ,  $1 \leq N \leq N_{\max}$ ,

$$X_N^{\text{nh}} \subset X, \quad \dim(X_N^{\text{nh}}) = N, \quad 1 \leq N \leq N_{\max}.$$

## Spaces & Bases – Lagrangian

Parameter samples:

$$S_N = \{\mu^1 \in \mathcal{D}, \dots, \mu^N \in \mathcal{D}\}, \quad 1 \leq N \leq N_{\max},$$

with

nested

$$S_1 \subset S_2 \subset \dots \subset S_{N_{\max}-1} \subset S_{N_{\max}} \subset \mathcal{D}.$$

Lagrangian reduced basis spaces:

hierarchical

$$W_N = \text{span}\left\{ \underbrace{u(\mu^n)}_{\text{“snapshots”}}, \quad 1 \leq n \leq N \right\}, \quad 1 \leq N \leq N_{\max},$$

with

$$W_1 \subset W_2 \subset \dots \subset W_{N_{\max}-1} \subset W_{N_{\max}} (\subset X).$$



## Spaces & Bases – Taylor & Hermite

Taylor reduced basis spaces: hierarchical

$$W_N^{\text{Taylor}} = \text{span}\{D_\sigma u(\mu), \forall \sigma \in I^{P,N-1}\}, \quad 1 \leq N \leq N_{\max},$$

field variable **and** sensitivity derivatives at **one point in  $\mathcal{D}$** .

Hermite reduced basis spaces: hierarchical

$$W_N^{\text{Hermite}} \text{ " = " } W_N^{\text{Lagrangian}} \cup W_N^{\text{Taylor}}$$

field variable **and** sensitivity derivatives at **several points in  $\mathcal{D}$**

Note: We will exclusively use Lagrangian RB spaces in this course.

## Spaces & Bases – Orthogonal Basis

Given  $\xi^n = u(\mu^n)$ ,  $1 \leq n \leq N_{\max}$  (Lagrange case) we construct the basis set  $\{\zeta^n\}$ ,  $1 \leq n \leq N_{\max}$ , from

### Gram-Schmidt Orthogonalization

$$\zeta^1 = \xi^1 / \|\xi^1\|_X;$$

for  $n = 2 : N_{\max}$

$$z^n = \xi^n - \sum_{m=1}^{n-1} (\xi^n, \zeta^m)_X \zeta^m;$$

$$\zeta^n = z^n / \|z^n\|_X;$$

end.

Note:  $(\zeta^n, \zeta^m)_X = \delta_{nm}$ ,  $1 \leq n, m \leq N_{\max}$

## Spaces & Bases – Orthogonal Basis

Given reduced basis space

$$X_N = \text{span}\{\zeta^n, 1 \leq n \leq N\}, \quad 1 \leq N \leq N_{\max}.$$

we can express any  $w_N \in X_N$  as

$$w_N = \sum_{n=1}^N w_{Nn} \zeta^n$$

for unique  $w_{Nn} \in \mathbb{R}$ ,  $1 \leq n \leq N$ .

Reduced basis “matrices”  $\underline{Z}_N \in \mathbb{R}^{N \times N}$ ,  $1 \leq N \leq N_{\max}$ :

$$\underline{Z}_N = [\zeta^1 \ \zeta^2 \ \dots \ \zeta^N], \quad 1 \leq N \leq N_{\max},$$

where, from orthogonality,  $\underline{Z}_{N_{\max}}^T \underline{X} \underline{Z}_{N_{\max}}^T = \underline{I}_{N_{\max}}$ , and  $\underline{I}_M$  is the Identity matrix in  $\mathbb{R}^{M \times M}$ .

## Galerkin Projection

Given  $\mu \in \mathcal{D} \subset \mathbb{R}^P$ , evaluate

optimality

$$s_N(\mu) = f(u_N(\mu); \mu)$$

where  $u_N(x; \mu) \in X_N \subset X^{\mathcal{N}}$  satisfies

$$a(u_N(\mu), v; \mu) = f(v; \mu), \quad \forall v \in X_N.$$

Note:

- ▶ Lagrangian reduced basis space:  $X_N = W_N$
- ▶ Solution  $u_N(\mu)$  unique (from coercivity, continuity, linear independence)

## Galerkin Optimality

### Proposition

For any  $\mu \in \mathcal{D}$ , we have

$$\begin{aligned} |||u(\mu) - u_N(\mu)|||_\mu &= \inf_{w_N \in X_N} |||u(\mu) - w_N(\mu)|||_\mu, \\ \|u(\mu) - u_N(\mu)\|_X &\leq \sqrt{\frac{\gamma^e(\mu)}{\alpha^e(\mu)}} \inf_{w_N \in X_N} \|u(\mu) - w_N(\mu)\|_X, \end{aligned}$$

and furthermore

$$\begin{aligned} s(\mu) - s_N(\mu) &= |||u(\mu) - u_N(\mu)|||_\mu^2 \\ &= \inf_{w_N \in X_N} |||u(\mu) - w_N(\mu)|||_\mu^2, \end{aligned}$$

as well as

$$0 \leq s(\mu) - s_N(\mu) \leq \gamma^e(\mu) \inf_{w_N \in X_N} \|u(\mu) - w_N(\mu)\|_X^2.$$

## Offline-Online Decomposition

We expand  $\mathbf{u}_N(\boldsymbol{\mu}) = \sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) \zeta^j$

and obtain

$$\mathbf{v} = \zeta^i, 1 \leq i \leq N$$

$$a(\mathbf{u}_N(\boldsymbol{\mu}), \mathbf{v}; \boldsymbol{\mu}) = f(\mathbf{v}; \boldsymbol{\mu})$$

$$\sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) a(\zeta^j, \zeta^i; \boldsymbol{\mu}) = f(\zeta^i; \boldsymbol{\mu})$$

$$\underbrace{\sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) \sum_{q=1}^{Q_a} \Theta_a^q(\boldsymbol{\mu}) \underbrace{a^q(\zeta^j, \zeta^i)}_{\text{OFFLINE: } O(N)}}_{\text{ONLINE: } O(Q_a N^2)} = \sum_{q=1}^{Q_f} \Theta_f^q(\boldsymbol{\mu}) \underbrace{f^q(\zeta^i)}_{\text{OFFLINE: } O(N)} \underbrace{\hspace{10em}}_{\text{ONLINE: } O(Q_f N)}$$

ONLINE:  $O(N^3)$

## Offline-Online Decomposition

We expand 
$$\mathbf{u}_N(\boldsymbol{\mu}) = \sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) \zeta^j$$

and obtain

$$\mathbf{v} = \zeta^i, 1 \leq i \leq N$$

$$a(\mathbf{u}_N(\boldsymbol{\mu}), \mathbf{v}; \boldsymbol{\mu}) = f(\mathbf{v}; \boldsymbol{\mu})$$

$$\sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) a(\zeta^j, \zeta^i; \boldsymbol{\mu}) = f(\zeta^i; \boldsymbol{\mu})$$

$$\underbrace{\sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) \sum_{q=1}^{Q_a} \Theta_a^q(\boldsymbol{\mu}) \underbrace{a^q(\zeta^j, \zeta^i)}_{\text{OFFLINE: } O(N)}}_{\text{ONLINE: } O(Q_a N^2)} = \sum_{q=1}^{Q_f} \Theta_f^q(\boldsymbol{\mu}) \underbrace{f^q(\zeta^i)}_{\text{OFFLINE: } O(N)} \underbrace{\quad}_{\text{ONLINE: } O(Q_f N)}$$

ONLINE:  $O(N^3)$

## Offline-Online Decomposition

We expand 
$$\mathbf{u}_N(\boldsymbol{\mu}) = \sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) \zeta^j$$

and obtain

$$\mathbf{v} = \zeta^i, 1 \leq i \leq N$$

$$a(\mathbf{u}_N(\boldsymbol{\mu}), \mathbf{v}; \boldsymbol{\mu}) = f(\mathbf{v}; \boldsymbol{\mu})$$

$$\sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) a(\zeta^j, \zeta^i; \boldsymbol{\mu}) = f(\zeta^i; \boldsymbol{\mu})$$

$$\underbrace{\sum_{j=1}^N \mathbf{u}_{Nj}(\boldsymbol{\mu}) \sum_{q=1}^{Q_a} \Theta_a^q(\boldsymbol{\mu}) \underbrace{a^q(\zeta^j, \zeta^i)}_{\text{OFFLINE: } O(\mathcal{N})}}_{\text{ONLINE: } O(Q_a N^2)} = \underbrace{\sum_{q=1}^{Q_f} \Theta_f^q(\boldsymbol{\mu}) \underbrace{f^q(\zeta^i)}_{\text{OFFLINE: } O(\mathcal{N})}}_{\text{ONLINE: } O(Q_f N)}$$

$$\underbrace{\hspace{15em}}_{\text{ONLINE: } O(N^3)}$$



## Offline-Online Decomposition

Given  $u_{Nj}(\mu)$ ,  $1 \leq j \leq N$ , we evaluate the output from

$$\begin{aligned}
 s_N(\mu) = f(u_N(\mu); \mu) &= \sum_{j=1}^N u_{Nj}(\mu) f(\zeta^j; \mu) \\
 &= \sum_{j=1}^N u_{Nj}(\mu) \underbrace{\sum_{q=1}^{Q_f} \Theta_f^q(\mu) \underbrace{f^q(\zeta^j)}_{\text{OFFLINE: } O(N)}}_{\text{ONLINE: } O(Q_f N)} \\
 &\underbrace{\hspace{10em}}_{\text{ONLINE: } O(N)}
 \end{aligned}$$

## Offline-Online Decomposition

Given  $u_{Nj}(\mu)$ ,  $1 \leq j \leq N$ , we evaluate the output from

$$\begin{aligned}
 s_N(\mu) = f(u_N(\mu); \mu) &= \sum_{j=1}^N u_{Nj}(\mu) f(\zeta^j; \mu) \\
 &= \underbrace{\sum_{j=1}^N u_{Nj}(\mu)}_{\text{ONLINE: } O(N)} \underbrace{\sum_{q=1}^{Q_f} \Theta_f^q(\mu)}_{\text{ONLINE: } O(Q_f N)} \underbrace{f^q(\zeta^j)}_{\text{OFFLINE: } O(N)}
 \end{aligned}$$

## Offline-Online Decomposition

Given  $u_{Nj}(\mu)$ ,  $1 \leq j \leq N$ , we evaluate the output from

$$\begin{aligned}
 s_N(\mu) = f(u_N(\mu); \mu) &= \sum_{j=1}^N u_{Nj}(\mu) f(\zeta^j; \mu) \\
 &= \underbrace{\sum_{j=1}^N u_{Nj}(\mu)}_{\text{ONLINE: } O(N)} \underbrace{\sum_{q=1}^{Q_f} \Theta_f^q(\mu) \underbrace{f^q(\zeta^j)}_{\text{OFFLINE: } O(N)}}_{\text{ONLINE: } O(Q_f N)}
 \end{aligned}$$

## Offline-Online Decomposition

Summary computational cost:

$$(Q = Q_a + Q_f)$$

*OFFLINE* — once, parameter *independent*

$$O(N_{\max} \mathcal{N}^\bullet) \quad + \quad O(Q N_{\max}^2 \mathcal{N}) \quad ;$$

solve for  $\zeta_n$                       form  $\mu$ -independent quantities

*ONLINE* — many times, parameter *dependent*

$$O(Q N^2) \quad + \quad O(N^3) \quad + \quad O(N) \quad \mu^{\text{new}}$$

form RB matrices                      solve for  $u_{N_j}(\mu)$                       evaluate output

Online cost is *independent* of  $\mathcal{N}$ .

## Algebraic Equations

Evaluation of RB Stiffness Matrix  $\underline{\mathbf{A}}_N \in \mathbb{R}^{N \times N}$ :

Parameter-independent matrices  $\underline{\mathbf{A}}_N^q \in \mathbb{R}^{N \times N}$ ,  $1 \leq q \leq Q_a$ :

$$\begin{aligned} \underline{\mathbf{A}}_{Nnm}^q &= a^q(\zeta^m, \zeta^n) \\ &= \sum_{i=1}^{\mathcal{N}} \sum_{j=1}^{\mathcal{N}} \zeta_i^m a^q(\varphi_i^{\mathcal{N}}, \varphi_j^{\mathcal{N}}) \zeta_j^n, \quad 1 \leq n, m \leq N, \end{aligned}$$

thus

$$\underline{\mathbf{A}}_N^q = \mathbf{Z}_N^T \underline{\mathbf{A}}^{\mathcal{N}q} \mathbf{Z}_N.$$

We finally assemble

$$\underline{\mathbf{A}}_N = \sum_{q=1}^{Q_a} \Theta_a^q(\mu) \underline{\mathbf{A}}_N^q.$$

## Algebraic Equations

Evaluation of RB Load/Source/Output Vector  $\underline{\mathbf{F}}_N \in \mathbb{R}^N$ :

Parameter-independent vectors  $\underline{\mathbb{F}}_N^q \in \mathbb{R}^N$ ,  $1 \leq q \leq Q_f$ :

$$\begin{aligned} \mathbb{F}_{Nn}^q &= f^q(\zeta^n) \\ &= \sum_{i=1}^{\mathcal{N}} \zeta_i^m f^q(\varphi_i^{\mathcal{N}}), \quad 1 \leq n \leq N, \end{aligned}$$

thus

$$\underline{\mathbb{F}}_N^q = \mathbb{Z}_N^T \underline{\mathbb{F}}^{\mathcal{N}q}.$$

We finally assemble

$$\underline{\mathbf{F}}_N = \sum_{q=1}^{Q_f} \Theta_f^q(\mu) \underline{\mathbb{F}}_N^q.$$

# Algebraic Equations

We expand  $u_N(\mu) = \sum_{j=1}^N u_{Nj}(\mu) \zeta^j$

and define the vector of RB coefficients

$$\underline{u}_N(\mu) = [u_{N1}(\mu) \ u_{N2}(\mu) \ \dots \ u_{NN}(\mu)] \in \mathbb{R}^N.$$

Then, given  $\mu \in \mathcal{D} \subset \mathbb{R}^P$ , we evaluate

$$s_N(\mu) = \underline{F}_N^T(\mu) \underline{u}_N(\mu)$$

where  $\underline{u}_N(\mu) \in \mathbb{R}^N$  satisfies

$$\underline{A}_N(\mu) \underline{u}_N(\mu) = \underline{F}_N(\mu).$$

## Algebraic Equations

### Proposition

The condition number of  $\underline{A}_N(\mu)$  is bounded from above by  $\gamma^e(\mu)/\alpha^e(\mu)$ , the ratio of the continuity and coercivity constants for the continuous problem.

Sketch of proof:

- ▶ Rayleigh Quotient

$$\frac{\underline{w}_N^T \underline{A}_N(\mu) \underline{w}_N}{\underline{w}_N^T \underline{w}_N}, \quad \forall \underline{w}_N \in \mathbb{R}^N$$

- ▶ Express

$$\underline{w}_N = \sum_{m=1}^N w_{N m} \zeta^m$$

- ▶ Invoke coercivity (resp. continuity) and orthogonality.