

MODEL ORDER REDUCTION

RIDUZIONE DEL MODELLO

BASI RIDOTTE / REDUCED BASIS

$$-\Delta u = f \quad \Omega_h \rightarrow V_h \ni u_h \leftarrow \begin{matrix} \text{FEM} \\ \text{FD} \end{matrix}$$

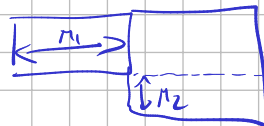
SISTEMA LINEARE

PARAMETRICO $\mu \in P \subset \mathbb{R}^P$

$M \rightarrow$ BOUNDARY CONDITIONS
 $M \rightarrow$ PARAMETRI FISICI
 $M \rightarrow$ GEOMETRIA

$$-\nabla \cdot (K \cdot \nabla u) = f$$

$K(x)$



- OBIETTIVO 1: RISOLVERE FEM/FD PER MOLTI PARAMETRI
- OBIETTIVO 2: TROVARE MODELLO SURROGATO DI FEM/FD CHE SIA PIÙ VELOCE DA RISOLVERE.

A) \Rightarrow POCO GRADI DI LIBERTÀ

B) \Rightarrow BUONA APPROSS. DEGLI FEM.

B) ✓ DOPO

\Rightarrow A) UNO SPAZIO $V_N = \langle \psi_i \rangle_{i=1}^N$

V_N è UNA BUONA APPROX.

FEM $a(\psi_i, \sum \psi_j u_j) = F(\psi_i) \quad \forall i=1, \dots, N_h$

IL NOSTRO PROB. PARAMETRICO

RON $a(\psi_i, \sum \psi_j \hat{u}_j) = F(\psi_i) \quad \forall i=1, \dots, N$

$N \ll N_h$
 \ll
 $10 \quad \ll \quad 10^6$

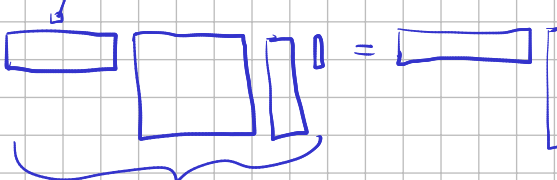

$$\psi_i(x) = \sum_{j=1}^{N_h} \psi_{i,j} \cdot \psi_j(x)$$

$$\underline{\psi}_i = \begin{pmatrix} \psi_{i,1} \\ \vdots \\ \psi_{i,N_h} \end{pmatrix} \in \mathbb{R}^{N_h}$$

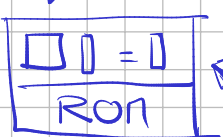
$$\sum_j \sum_{i,j} a(\psi_{i,j} \cdot \psi_j, \sum_k \sum_{\ell} \psi_{k,\ell} \cdot \psi_{\ell} \cdot \hat{u}_k) = F(\sum_j \psi_{i,j} \cdot \psi_j)$$

$$\forall i=1, \dots, N \quad \sum_{j=1}^{N_h} \psi_{i,j} \overbrace{a(\psi_j, \psi_{\ell})}^N \sum_{\ell=1}^{N_h} \sum_{k=1}^N \psi_{k,\ell} \cdot \hat{u}_k = \sum_j \psi_{i,j} \overbrace{F(\psi_j)}^{FEN}$$

$$\underline{R} \in \mathbb{R}^{N_h \times N} \quad R = \begin{bmatrix} \overbrace{\psi_1, \psi_2, \dots, \psi_N}^N \end{bmatrix} \begin{matrix} \downarrow N_h \end{matrix} \quad R = \begin{bmatrix} \end{bmatrix}$$

$$R^T A \cdot R \cdot \hat{u} = R^T \cdot \underline{F}$$



$$\hat{A} = R^T A R \quad \hat{F} = R^T \cdot F$$

$$\hat{A} \hat{u} = \hat{F}$$


• $A(\mu) \quad F(\mu) \leadsto R^T A(\mu) R$

• SE C'È UNA DIPENDENZA AFFINE DI A E F DAL PARAMETRO μ

$$A(\mu) = \sum_{k=1}^{M_A} \underbrace{\Theta_k^A(\mu)}_{\uparrow R} \cdot \underbrace{A_k}_{\downarrow \text{INDIP. DA } \mu}$$

$$F(\mu) = \sum_{k=1}^{M_F} \underbrace{\Theta_k^F(\mu)}_{\uparrow R} \cdot \underbrace{F_k}_{\downarrow \text{INDIP. DA } \mu}$$

$$R^T A(\mu) R = \sum_{k=1}^{M_A} R^T \Theta_k^A(\mu) A_k \cdot R = \sum_{k=1}^{M_A} \Theta_k^A(\mu) \underbrace{R^T A_k R}_{\hat{A}_k} = \sum_{k=1}^{M_A} \Theta_k^A(\mu) \underbrace{\hat{A}_k}_{\in \mathbb{R}^{N \times N}}$$

$$R^T F(\mu) = \sum_{k=1}^{M_F} \Theta_k^F(\mu) \cdot \underbrace{R^T F_k}_{\hat{F}_k} = \sum_{k=1}^{M_F} \Theta_k^F(\mu) \cdot \underbrace{\hat{F}_k}_{\uparrow \mathbb{R}^N}$$

• CONE SCEGLIERE V_N E LA SUA BASE $\{\psi_i\}_{i=1}^N \Rightarrow R$ MATRICE DI PROIEZIONE?

\hookrightarrow 1. GENERARE DEGLI SNAPSHOTS $\{u(\mu_1), \dots, u(\mu_{M_{\text{TRAIN}}})\}$ DI PROIEZIONE

POD PROPER ORTHOGONAL DECOMPOSITION

SVD SINGULAR VALUE DECOMPOSITION

$$S = [u(\mu_1), \dots, u(\mu_{M_{\text{TRAIN}}})] \in \mathbb{R}^{N_h \times M_{\text{TRAIN}}}$$

$$SVD(S) = U \Sigma V^T = S$$

$$U, V \text{ ORTOGONALI} \quad U^T U = I \quad V^T V = I$$

$$U \in \mathbb{R}^{N_h \times N_h}$$

$$\Sigma \in \mathbb{R}^{N_h \times m_{TRAIN}}$$

$$V \in \mathbb{R}^{m_{TRAIN} \times m_{TRAIN}}$$

$$\Sigma \text{ \u00c9 DIAGONALE} \quad \Sigma > 0$$

$$S \approx \hat{U} \hat{\Sigma} \hat{V}^T$$

$$\hat{\Sigma} \in \mathbb{R}^{N_h \times N_h}$$

$$R \leftarrow \hat{U} \in \mathbb{R}^{N_h \times N_h}$$

$$\hat{V} \in \mathbb{R}^{N_h \times m_{TRAIN}}$$

\hat{U} HA BUONA

PARTE DELLE INFORMAZIONI DI S

$$S - \hat{U} \hat{\Sigma} \hat{V}^T = \text{PICCOLO}$$

SPAZIO RIDOTTO



$$(\Theta_0^q(N) A_0 + \overset{\text{NON BORRO}}{\Theta_1^q A_1}) q = \Theta_0^f \cdot F_0$$

DIRETTORE \Rightarrow

$$A_0[i:] = \begin{bmatrix} 0 & \dots & 0 & 1 & 0 & \dots & 0 \\ \vdots & & \vdots & & \vdots & & \vdots \end{bmatrix}$$

$\forall i$ BORRO DIRETTORE

$$F_0[i] = [0] \leftarrow \text{onecello}$$

$$\underbrace{\Theta_0^q(N)}_{\neq 0} \cdot \mathbb{1} \cdot U_i = 0 \quad \Rightarrow \quad U_i = 0 \quad \text{BENE COSÌ}$$

DISINTEGRO \Rightarrow PIÙ COMPLICATO

$$(R^T A_k R)_{ij} =$$

$$\cancel{(R^T (A_k R))_{ij}}$$

$$a_k(\psi_i, \psi_j)$$

$$\sum_{\ell, z} (R^T)_{i\ell} (A_k)_{\ell z} (R)_{zj}$$

$$(R_i^T) A \cdot R_{:j}$$

$$R_i^T A_k R_j \quad \begin{matrix} \downarrow & \downarrow \\ \mathbb{R}^{N_h} & \mathbb{R}^{N_h \times N_h} \end{matrix} \quad \begin{matrix} \in \mathbb{R}^{N_h} \\ \forall i, j = 1, \dots, N \end{matrix}$$

$$R_i^T \cdot (A_k R_j) = \hat{A}_{ij}$$

- LINEARE PROBLEM
- DECOMPOSIZIONE AFFINE NEI PARAMETRI
- VALORI SINGOLARI DECA SONO ESPONENZIALMENTE

$$\bullet \quad \partial_{\underline{t}} u - \Delta_{\underline{t}} u = 0$$

$$\bullet \quad \partial_{\underline{t}} u + a \partial_x u = 0$$