

NMPDE/ATSC 2025

Lecture 10

(Q. 4.5.2)

$$A \cup = F$$

Assuming A is SPD associated to the

$$(A \Sigma, \Sigma) = A (\Sigma_n, \Sigma_n)$$

where $v_n = \sum_i v_i \varphi_i$; $\frac{v}{\uparrow} = (v_i)_{i \in \text{coeffs}}$

$$\chi_{SP}(A) = \frac{\lambda_{\max}(A)}{\lambda_{\min}(A)}$$

Lemma: Let Z_n quasi-uniform, shape-regular. Then

$$\exists c_1, c_2 > 0 :$$

$$c_1 h^d |\underline{w}|^2 \leq \|w_h\|_0^2 \leq c_2 h^d |\underline{w}|^2 \quad \forall w_h \in V_h^k$$

Theorem: Some assumptions $\Rightarrow \chi_{sp}(A) = O(\bar{n}^2)$

$$\text{Proof: } \frac{(A\underline{v}, \underline{v})}{|\underline{v}|^2} = \frac{f(A\underline{v}_h, \underline{v}_h)}{|\underline{v}|^2} \leq \gamma \frac{\|\underline{v}_h\|_1^2}{|\underline{v}|^2}$$

$$R_A(v) = \underbrace{1 -}_{\text{Reileigh quotient}}$$

continuity

$$\|\boldsymbol{\nu}_h\|_1^2 = \|\boldsymbol{\nu}_h\|_0^2 + |\boldsymbol{\nu}_h|_1^2$$

Proposition (H^1-L^2 Inverse inequality): Some
obs. $\Rightarrow \exists C_{\text{inv}} > 0$:

$$|\boldsymbol{\nu}_h|_1^2 \leq C_{\text{inv}} h^{-2} \|\boldsymbol{\nu}_h\|_0^2 \quad \forall \boldsymbol{\nu}_h \in V_h^k$$

$$\leq (1 + C_{\text{inv}} h^{-2}) \|\boldsymbol{\nu}_h\|_0^2$$

$$R_A(\boldsymbol{\nu}) \leq \gamma (1 + C_{\text{inv}} h^{-2}) \frac{\|\boldsymbol{\nu}_h\|_0^2}{\|\boldsymbol{\nu}\|^2} \leq C_2 \gamma h^d (1 + C_{\text{inv}} h^{-2})$$

$$R_A(\boldsymbol{\nu}) = \frac{\mathcal{F}(\boldsymbol{\nu}_h, \boldsymbol{\nu}_h)}{\|\boldsymbol{\nu}\|^2} \geq d_0 \frac{\|\boldsymbol{\nu}_h\|_1^2}{\|\boldsymbol{\nu}\|^2} \geq d_0 \frac{\|\boldsymbol{\nu}_h\|_0^2}{\|\boldsymbol{\nu}\|^2} \geq C_1 d_0 h^d$$

coercivity

$$\frac{C_1 d_0 h^d}{C_2 \gamma h^d} \leq R_A(\boldsymbol{\nu}) \leq C_2 \gamma h^d (1 + C_{\text{inv}} h^{-2})$$

In particular $\lambda_{\min}(A) \leq \lambda_{\max}(A) \leq$

$$\Rightarrow \chi_{\text{sp}}(A) = \frac{\lambda_{\max}(A)}{\lambda_{\min}(A)} \leq \frac{C_2 \gamma}{C_1} \frac{2}{2} (1 + C_{\text{inv}} h^{-2}) = O(h^{-2})$$

Remark: - independent from space dim, conditioning linked to discretization parameter h (more fine is the mesh)

- As the space dim grows, expect larger systems ...

(Q. Ch7)

SOLUTION OF $AU = F$

Let $m = \text{dim. of the system}$ ($= \# \text{DoF}$ of FEM)

① DIRECT SOLVERS

e.g. LU decomposition (Gauss elimination) $\mathcal{O}(\frac{2}{3}m^3)$

Advantages:

- once the LU decomposition is computed, can solve for different F
- incomplete LU versions available
- if A Symmetric \Rightarrow Cholesky method $\mathcal{O}(\frac{1}{3}m^3)$
- for Banded A , e.g. stemming from FE discretisations cost reduces to $\mathcal{O}(m^2)$
- tridiagonal $A \Rightarrow$ Thomas algorithm $\mathcal{O}(m)$

Rule of Thumb: use direct methods for $d \leq 2$.

② Iterative solvers: produce a sequence of vectors $\{v^{(n)}\}_n \xrightarrow{n} v$ sol. of $AU=F$

- decompose $A = P - H$

- given an initial guess $v^{(0)}$, iterate

$$Pv^{(n+1)} = Hv^{(n)} + F$$

requiring many solutions w.r.t. matrix P

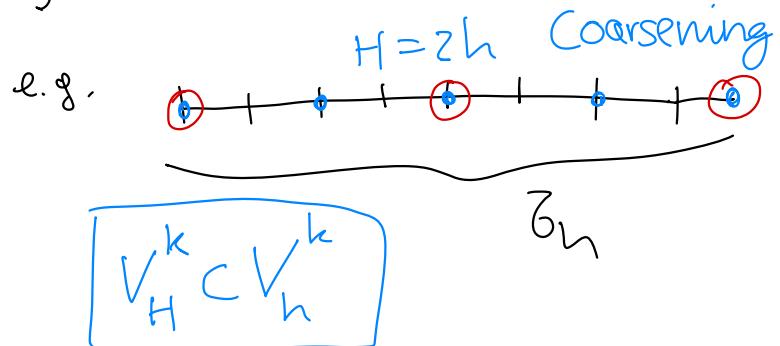
- \Rightarrow need
 - P invertible
 - P easy to invert

Pre conditioner

Examples:

- A symmetric : Conjugate gradient
- A general : Krylov subspaces
(GMRES, bCG, BiCG Stab) for FEM sys
- Geometric Multigrid $\rightsquigarrow O(m)$
(GMG)
 - To solve over given mesh \mathcal{T}_h ,

GMG exploits a sequence of hierarchical meshes coarsenings of \mathcal{T}_h



The idea is to stop iteration for the solution of $A_h U_h = F_h$ before convergence and correct it by a (cheaper ?) solve of $A_H \delta_H = F_H$

An idea of the GMG algorithm steps

close
two meshes
 $\mathcal{T}_h, \mathcal{T}_H$
V-cycle

① $U_h^{(0)} = S_h(U_h^{(l-1)}, F_h)$ starting from some $U_h^{(0)}$
l = 1, ..., m_1

② $r_h = F_h - A_h U_h^{(m_1)}$

③ $r_H = I_h^H r_h$ restrict residual

④ $A_H \delta_H = r_H$ some fine-to-coarse operator coarse solve

$$(5) \quad U_h^{(m_1+1)} = U_h^{(m_1)} + I_H^h S_H \quad \begin{matrix} \text{coarse-grid} \\ \text{correction} \end{matrix}$$

↓

coarse-to-fine op.

$$(6) \quad U_h^{(l)} = S_h (U_h^{(l-1)}, F_h) \quad l = m_1 + 1, \dots, m_1 + m_2 + 1$$

Iterative methods require many matrix-vector multiplications \Rightarrow crucial to do such multiplications efficiently

One way to do this is by
MATRIX FREE approaches