

Iterative Methods - April 12, 2006, 3

Note Title

4/12/2006

$$A = \begin{pmatrix} A_1 & B_1 & & \\ B_1^T & \ddots & & \\ & \ddots & B_{n-1} & \\ & & B_{n-1}^T & A_n \end{pmatrix}$$

$$M = \begin{pmatrix} A_1 & & & \\ & \ddots & & 0 \\ & & \ddots & \\ 0 & & & A_n \end{pmatrix}; \rho(M^{-1}N) = ?$$

Point Jacobi: $\rho(M^{-1}N) \leq \max_i \sum_{j \in E_i} \left| \frac{a_{ij}}{a_{ii}} \right|$

$$\underline{M^{-1} N} \underline{z} = \lambda \underline{z}, \quad \underline{z} = \begin{pmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{pmatrix}$$

$$- A_j^{-1} B_{j-1}^T z_{j-1} - A_j^{-1} B_{j+1} z_{j+1} = \lambda z_j$$

$$\max |\lambda| \leq \|A_j^{-1} B_{j-1}^T\| \frac{\|z_{j-1}\|}{\|z_j\|} + \|A_j^{-1} B_{j+1}\| \cdot \frac{\|z_{j+1}\|}{\|z_j\|}$$

$$\leq \max_j \left(\|A_j^{-1} B_{j-1}^T\| + \|A_j^{-1} B_{j+1}\| \right)$$

$$A = \begin{pmatrix} I & F \\ F^T & H \end{pmatrix}$$

$$B_j = - \begin{pmatrix} 0 & F \\ F^T & 0 \end{pmatrix}$$

$$\begin{aligned} B_{GS} &= - (D+L)^{-1} U = - \begin{pmatrix} I & 0 \\ -F^T & I \end{pmatrix} \begin{pmatrix} 0 & F \\ 0 & 0 \end{pmatrix} \\ &= - \begin{pmatrix} I & 0 \\ -F^T & I \end{pmatrix} \begin{pmatrix} 0 & F \\ 0 & 0 \end{pmatrix} \\ &= + \begin{pmatrix} 0 & -F \\ 0 & F^T F \end{pmatrix} \end{aligned}$$

$$\lambda(B_{ES}) = 0 \cup \lambda(F^T F)$$

$$\lambda(B_J) = \lambda \begin{pmatrix} 0 & F \\ F^T & 0 \end{pmatrix}$$

$$B_J^2 = \begin{pmatrix} 0 & F \\ F^T & 0 \end{pmatrix} \begin{pmatrix} 0 & F \\ F^T & 0 \end{pmatrix} = \begin{pmatrix} FF^T & 0 \\ 0 & F^T F \end{pmatrix}$$

$$\lambda(AB) = \lambda(BA)$$

$$\left[\begin{array}{cc} A_{100 \times 10} & A^T A_{10 \times 10} \\ & A A^T_{100 \times 100} \end{array} \right] \quad \lambda(AA^T) = \lambda(A^T A) \cup \{0\}$$

$$\rho(B_{ES}) = \rho^2(B_J)$$

$$A = \begin{pmatrix} 2 & -1 & & 0 \\ -1 & 2 & & \\ & \ddots & \ddots & \\ 0 & -1 & & 2 \end{pmatrix}_{N \times N}$$

Jacobi

$$\lambda(M^{-1}N) = \cos \frac{\pi}{N+1}, \quad \frac{1}{N+1} = h$$

$$= 1 - \frac{\pi^2 h^2}{2} + \mathcal{O}(h^4)$$

$$\rho(G-S) = 1 - \pi^2 h^2 + \mathcal{O}(h^4)$$

$$\frac{\|c^k\|}{\|c^0\|} \leq (\cos \pi h)^k \leq \varepsilon, \quad k = \frac{-\ln \varepsilon}{-\ln \cos \pi h}$$

$$\sim -\ln \varepsilon / \pi^2 h^2 / 2$$

- $\ln p(M^{-1}N)$: rate of convergence

of iterations $\sim h^{-2}$

ACCELERATION

\tilde{x}^0 : given

$$\tilde{x}^{l+1} = \tilde{x}^l + \alpha_{l+1} \tilde{r}^l$$

$$\tilde{r}^l = \tilde{b} - A \tilde{x}^l$$

$$\underline{A = A^T \text{ p.d.}}$$

$$\tilde{e}^l = \tilde{x}^l - \tilde{x}$$

$$\tilde{x}^{l+1} = \tilde{x}^l + \alpha_{l+1} \tilde{r}^l = \tilde{x}^l + \alpha_{l+1} (\tilde{b} - A \tilde{x}^l)$$

$$\tilde{x} = \tilde{x}$$

$$\tilde{e}^{l+1} = \tilde{e}^l + \alpha_{l+1} (A (\tilde{x} - \tilde{x}^l))$$

$$= (I - \alpha_{l+1} A) \tilde{e}^l$$

$$\tilde{e}^1 = (I - \alpha_1 A) \tilde{e}^0$$

$$\tilde{e}^2 = (I - \alpha_2 A) (I - \alpha_1 A) \tilde{e}^0$$

$$e^k = (I - \alpha_k A)(I - \alpha_{k-1} A) \dots (I - \alpha_1 A) \tilde{e}^0$$

$$\equiv P_k(A) \tilde{e}^0$$

$$P_k(\lambda) = (1 - \alpha_k \lambda)(1 - \alpha_{k-1} \lambda) \dots (1 - \alpha_1 \lambda)$$

$$P_k(0) = 1$$

$$A = Q \Lambda Q^T, \quad \tilde{e}^k = Q P_k(\Lambda) Q^T \tilde{e}^0$$

$$\|\tilde{e}^k\|_2 \leq \|Q P_k(\Lambda) Q^T\|_2 \|\tilde{e}^0\|_2$$

$$= \|P_k(\Lambda)\|_2 \cdot \|\tilde{e}^0\|_2$$

$$\min_{\{x_k\}} \max_{\{\lambda_i\}} |p_k(\lambda)| = \rho$$

$$P_k(0) = 1$$

$k = n$. Cayley - Hamilton Theorem.

$$p(\lambda) = \sum_{i=0}^n \beta_i \lambda^i = 0$$

$$p(A) = \sum \beta_i A^i = 0$$

$$\sum_{i=0}^n \beta_i \lambda^i \approx \prod_{i=1}^n (\lambda - \lambda_i) = 0$$

$$\prod \left(1 - \frac{\lambda}{\lambda_i}\right) = 0$$

$$\alpha_i = \frac{1}{\lambda_i}$$

Terrible!!

$$\underline{k=1}$$

$$\underline{C}^k = (I - \alpha A)^k \underline{C}^0$$

$$p_k(\lambda) = (1 - \alpha\lambda)^k$$

$$\min \max |1 - \alpha\lambda|$$

$$a \leq \lambda \leq b$$

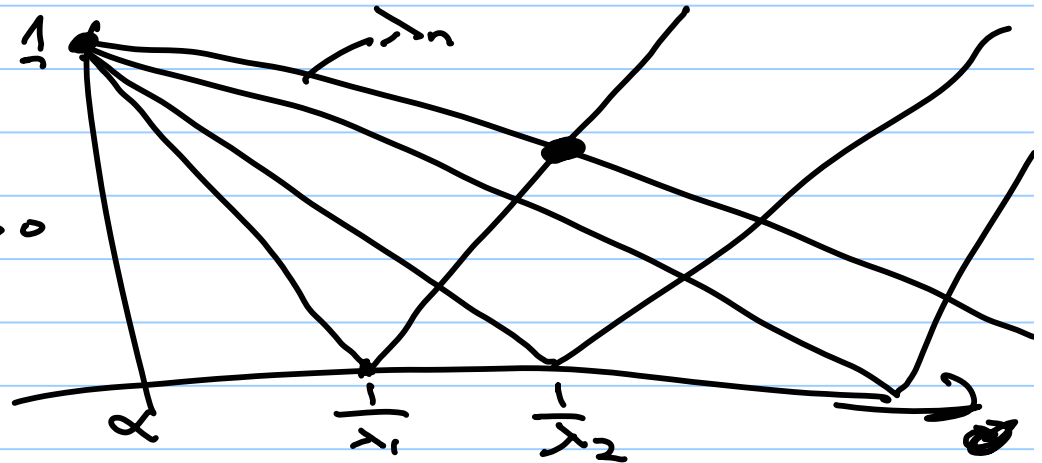
$$p_1(\lambda) = (1 - \alpha\lambda)$$

$$p_1^{(1)}(\lambda_1)$$

$$p_1^{(2)}(\lambda_2)$$

$$p_1(\lambda)$$

$$\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n > 0$$



Optimal

$$1 - \hat{\alpha} \lambda_n = - (1 - \hat{\alpha} \lambda_1)$$

$$\hat{\alpha} (\lambda_1 + \lambda_n) = 2$$

$$\hat{\alpha} = \frac{2}{\lambda_1 + \lambda_n}$$

$$\rho = |1 - \hat{\alpha} \lambda_n| = \frac{\lambda_1 - \lambda_n}{\lambda_1 + \lambda_n} = \frac{k-1}{k+1}$$

$$k = \lambda_1 / \lambda_n$$

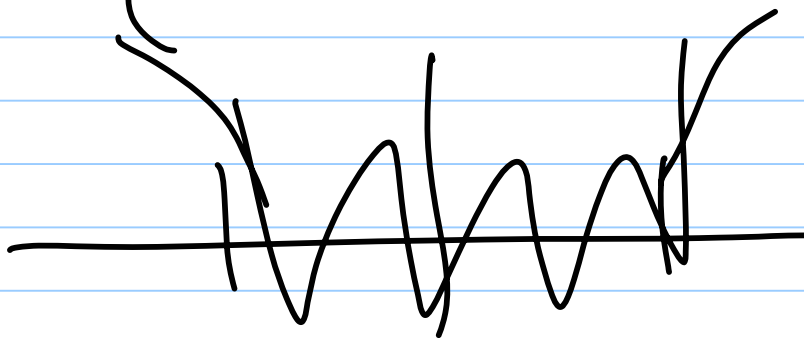
k : parameter

$$\begin{aligned} \rho(A^{n-1}N) &= \rho(I - \hat{\alpha}A) \\ &= \lambda_{\max}(I - \hat{\alpha}A) \end{aligned}$$

Chebyshev

$$a \leq x_i \leq b$$

$$T_k(\mu) = \begin{cases} \cos(k \cos^{-1} \mu) & |\mu| \leq 1 \\ \cosh(k \cosh^{-1} \mu) & \mu \geq 1 \end{cases}$$



$$\hat{P}_k(\lambda) = \frac{T_k \left(\frac{2\lambda - (a+b)}{b-a} \right)}{T_k \left(\frac{a+b}{b-a} \right)}$$

$$\downarrow$$
$$\mu_k = \frac{a+b}{2} + \frac{b-a}{2} \cos \left(\frac{2k-1}{k} \cdot \frac{\pi}{2} \right) \quad k=1,2,\dots,k$$

$$\sigma_k = 1/\mu_k$$

$$|\hat{p}_r(\lambda)| \leq 2 \left(\frac{\sqrt{k} - 1}{\sqrt{k} + 1} \right)^k$$

$$k = \frac{\lambda_1}{\lambda_n}$$

Louisy Method

Why?

- 1) You precise estimates of eigenvalues.
- 2) k is determined & fixed.
- 3) Ordering is important.

Lebedev & Finogenov

$$\tilde{r}_k = (I - \alpha A) \tilde{r}_{k-1}$$

$$\tilde{x}^{k+1} = \tilde{x}^1 + \alpha \tilde{r}^k$$

$$\tilde{e}^{k+1} = (I - \alpha A) \tilde{e}^k$$

$$\tilde{e}^k = \tilde{x} - \tilde{x}^k$$
$$A \tilde{e}^k = \tilde{b} - A \tilde{x}^k = \tilde{r}^k$$

$$A \tilde{e}^{k+1} = (I - \alpha A) A \tilde{e}^k$$

$$\tilde{r}^{k+1} = (I - \alpha A) \tilde{r}^k = (I - \alpha A)^{k+1} \tilde{r}^0$$

$$A z_i = \lambda_i z_i \quad i=1, 2, \dots, n.$$

$$r^0 = \sum_{i=1}^n \beta_i z_i$$

$$r^k = (I - \alpha A)^k r^0 = (I - \alpha A)^k \sum_{i=1}^n \beta_i z_i$$

$$= \sum \beta_i (1 - \alpha \lambda_i)^k z_i$$

$$\begin{pmatrix} r^k \\ r^0 \end{pmatrix} = \sum_{i=1}^n \beta_i^2 (1 - \alpha \lambda_i)^k$$

$$\dots = \sum_{i=1}^n \beta_i^2 \alpha_i^k = \mu_k$$



$$\mu_k = \int \alpha^k dF(\beta)$$

$$\tilde{r}^{k+1} = \tilde{r}^k - \alpha_{k+1} A \tilde{r}^k$$

$$\min_{\alpha_{k+1}} (\tilde{r}^{k+1}, A^{-1} \tilde{r}^{k+1})$$

$$F(\alpha_{k+1}) = (\tilde{r}^{k+1}, A^{-1} \tilde{r}^{k+1})$$

$$= (\tilde{r}^k, A^{-1} \tilde{r}^k) - 2 \alpha_{k+1} (\tilde{r}^k, \tilde{r}^k) + \alpha_{k+1}^2 (A \tilde{r}^k, \tilde{r}^k)$$

$$\tilde{\alpha}_{k+1} = \frac{(\tilde{r}^k, \tilde{r}^k)}{(A \tilde{r}^k, \tilde{r}^k)}$$

\tilde{x}^0 : given

$$\tilde{x}^{k+1} = \tilde{x}^k + \alpha_{k+1} \tilde{r}^k,$$

$$\tilde{r}^k = \tilde{b} - A \tilde{x}^k$$

$$\alpha_{k+1} = \frac{(\tilde{r}^k, \tilde{r}^k)}{(A \tilde{r}^k, \tilde{r}^k)}$$
