

Lecture Note 5

**Iterative Linear System Solution
Methods (SOR, CCSI)**

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Classical Iterative Solution Methods

- Block (Line) Jacobi with Natural ordering

$$-L_i\phi_{i-1} + D_i\phi_i - U_i\phi_{i+1} = b_i$$

$$D_i\phi_i^{(l)} = b_i + L_i\phi_{i-1}^{(l-1)} + U_i\phi_{i+1}^{(l-1)}$$

$$\begin{bmatrix} \begin{bmatrix} -L_i \\ \diagdown \end{bmatrix} & \begin{bmatrix} \diagup \\ D_i \\ \diagdown \end{bmatrix} & \begin{bmatrix} \diagdown \\ -U_i \\ \end{bmatrix} \end{bmatrix} \begin{bmatrix} \phi_{i-1} \\ \phi_i \\ \phi_{i+1} \end{bmatrix} = \begin{bmatrix} \\ b_i \\ \end{bmatrix}$$

– Need LU factorization, then forward and backward substitution

- Block (Line) Gauss-Seidel with Natural ordering

$$-L_i\phi_{i-1} + D_i\phi_i - U_i\phi_{i+1} = b_i$$

$$D_i\phi_i^{(l)} = b_i + L_i\phi_{i-1}^{(l)} + U_i\phi_{i+1}^{(l-1)}$$

- SOR

$$\phi_i^{(l)} = \omega\phi_{i,GS}^{(l)} + (1 - \omega)\phi_i^{(l-1)}$$

Iterative Scheme and Convergence Condition

- Iterative scheme

$$A\phi = b$$

$$\text{Let } A = D - L - U = M - N$$

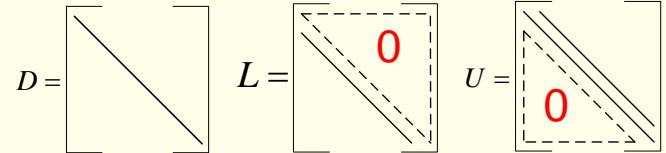
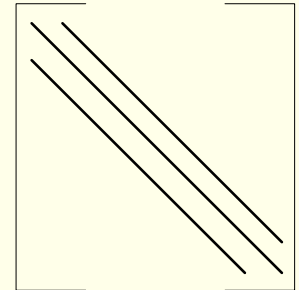
$$(M - N)\phi = b$$

$$M\phi = b + N\phi$$

$$M\phi^{(l)} = b + N\phi^{(l-1)} \quad : \text{ Iterative scheme}$$

$$\begin{cases} M = D \text{ and } N = L + U & : \text{ Jacobi} \\ M = D - L \text{ and } N = U & : \text{ G-S} \end{cases}$$

$$A = [a_{ij}] =$$



- Convergence criteria

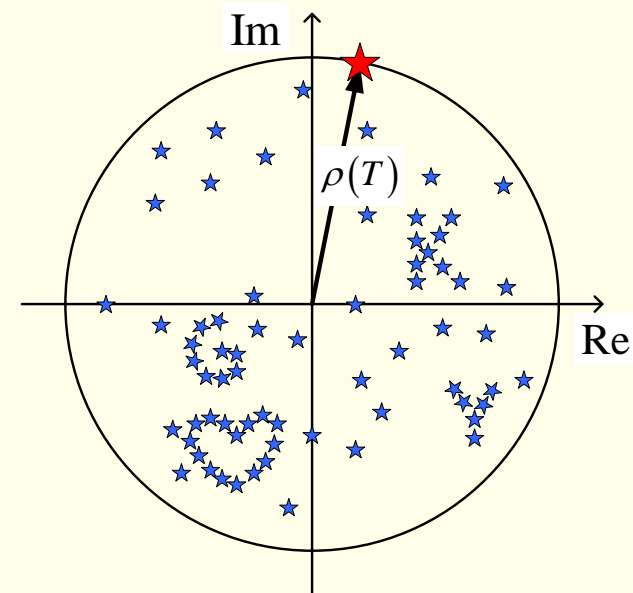
$$\begin{cases} M\phi^{(l)} = b + N\phi^{(l-1)} \\ M\phi^* = b + N\phi^* \end{cases} \quad \phi^* : \text{ Exact solution}$$

$$Me^{(l)} = Ne^{(l-1)}$$

$$e^{(l)} = M^{-1}Ne^{(l-1)} \quad \otimes T = M^{-1}N \quad : \text{ Iteration Matrix}$$

$$e^{(l)} = Te^{(l-1)} = T^2e^{(l-2)} = T^le^{(0)} = T^l \sum_i c_i u_i = \sum_i \lambda_i^l c_i u_i$$

$$\text{Converge if } |\rho(T)| = \left| \max(\lambda_i) \right| < 1 \quad \otimes \rho(T) : \text{ Spectral Radius}$$



Iterative Method in Matrix Form

- Jacobi

$$A\phi = b$$

$$\text{Let } A = D - (L + U) = M - N \quad (M = D \text{ and } N = L + U)$$

$$M\phi = b + N\phi$$

$$D\phi^{(l)} = b + (L + U)\phi^{(l-1)}$$

$$\phi^{(l)} = D^{-1} \left(b + (L + U)\phi^{(l-1)} \right)$$

$$\begin{aligned} \otimes T &= M^{-1}N \\ &= D^{-1}(L + U) \end{aligned}$$

- Gauss-Seidel

$$A\phi = b$$

$$\text{Let } A = (D - L) - U = M - N \quad (M = D - L \text{ and } N = U)$$

$$M\phi = b + N\phi$$

$$(D - L)\phi = b + U\phi$$

$$D\phi^{(l)} = b + L\phi^{(l)} + U\phi^{(l-1)}$$

$$\phi^{(l)} = D^{-1} \left(b + L\phi^{(l)} + U\phi^{(l-1)} \right)$$

$$\begin{aligned} \otimes T &= M^{-1}N \\ &= (D - L)^{-1}U \end{aligned}$$

Iterative Method in Matrix Form

- SOR

$$A\phi = b$$

$$\omega A\phi = \omega b$$

$$\text{Let } \omega A = \omega(D - L - U) = \omega D - \omega L - \omega U + D - D = \underbrace{D - \omega L}_M - \underbrace{[(1 - \omega D) + \omega U]}_N$$

$$(M - N)\phi = \omega b$$

$$M\phi = \omega b + N\phi$$

$$M\phi^{(l)} = \omega b + N\phi^{(l-1)}$$

$$(D - \omega L)\phi^{(l)} = \omega b + ((1 - \omega)D + \omega U)\phi^{(l-1)}$$

$$\therefore T = M^{-1}N$$

$$= (D - \omega L)^{-1} [(1 - \omega D) + \omega U]$$

$$D\phi^{(l)} = \omega(b + L\phi^{(l)} + U\phi^{(l-1)}) + (1 - \omega)D\phi^{(l-1)}$$

$$\phi^{(l)} = \omega\phi_{G.S}^{(l)} + (1 - \omega)\phi^{(l-1)}$$

– How to find a proper ω for $\min \rho(T)$?

→ can be answered by knowing the property of Jacobi iteration Matrix

Properties of Jacobi Iteration Matrix

- Properties of Jacobi Iteration Matrix

- 1) Jacobi Iteration Matrix with various Iteration Matrix Structure

- ① Natural ordering (2D)

$$\text{Let } A = D - (L + U) = M - N \quad (M = D \text{ and } N = L + U)$$

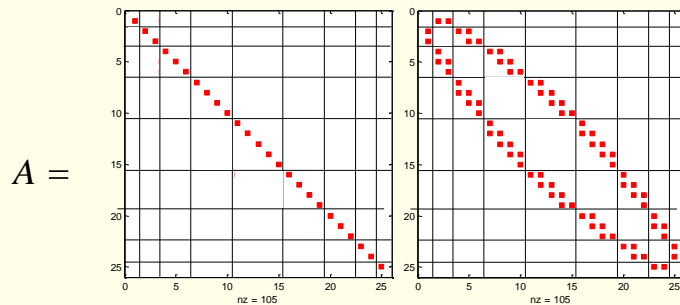
$$\begin{bmatrix} \boxed{\text{diag}} & \boxed{\text{diag}} & \\ \boxed{\text{diag}} & \boxed{\text{diag}} & \boxed{\text{diag}} \\ & \boxed{\text{diag}} & \boxed{\text{diag}} \end{bmatrix} = \begin{bmatrix} \boxed{\text{diag}} & & \\ & \boxed{\text{diag}} & \\ & & \boxed{\text{diag}} \end{bmatrix} + \begin{bmatrix} \boxed{\text{diag}} & \boxed{\text{diag}} & \\ \boxed{\text{diag}} & \boxed{\text{diag}} & \boxed{\text{diag}} \\ & \boxed{\text{diag}} & \boxed{\text{diag}} \end{bmatrix}$$

$$T = M^{-1}N = D^{-1}(L + U) = \begin{bmatrix} \boxed{\text{diag}} & \boxed{\text{diag}} & \\ \boxed{\text{diag}} & \boxed{\text{diag}} & \boxed{\text{diag}} \\ & \boxed{\text{diag}} & \boxed{\text{diag}} \end{bmatrix}$$

Properties of Jacobi Iteration Matrix

Ⓛ Point Jacobi Iteration Matrix with C-M

$$\text{Let } A = D - (L + U) = M - N \quad (M = D \text{ and } N = L + U)$$



$$T = D^{-1}(L + U) = \begin{bmatrix} D_1^{-1} & 0 & & 0 \\ 0 & D_2^{-1} & & 0 \\ & 0 & D_3^{-1} & 0 \\ & & 0 & D_4^{-1} & 0 \\ 0 & & & 0 & D_5^{-1} \end{bmatrix} \begin{bmatrix} 0 & U_1 & & 0 \\ L_2 & 0 & U_2 & \\ & L_3 & 0 & U_3 \\ & & L_4 & 0 & U_4 \\ 0 & & & L_5 & 0 \end{bmatrix} = \begin{bmatrix} 0 & D_1^{-1}U_1 & & 0 \\ D_2^{-1}L_2 & 0 & D_2^{-1}U_2 & \\ & D_3^{-1}L_3 & 0 & D_3^{-1}U_3 \\ & & D_4^{-1}L_4 & 0 & D_4^{-1}U_4 \\ 0 & & & D_5^{-1}L_5 & 0 \end{bmatrix}$$

– Diagonal blocks are all zero : consistently ordered

Properties of Jacobi Iteration Matrix

⊖ Red-Black

$$\text{Let } A = D - (L + U) = M - N \quad (M = D \text{ and } N = L + U)$$

$$A = \begin{bmatrix} D_1 & U_1 \\ L_2 & D_2 \end{bmatrix} = \begin{bmatrix} D_1 & 0 \\ 0 & D_2 \end{bmatrix} + \begin{bmatrix} 0 & U_1 \\ L_2 & 0 \end{bmatrix}$$

$$T = D^{-1}(L + U) = \begin{bmatrix} D_1^{-1} & 0 \\ 0 & D_2^{-1} \end{bmatrix} \begin{bmatrix} 0 & U_1 \\ L_2 & 0 \end{bmatrix} = \begin{bmatrix} 0 & D_1^{-1}U_1 \\ D_2^{-1}L_2 & 0 \end{bmatrix}$$

♠ Discussion

- Depending on the ordering scheme, the diagonal blocks of the Jacobi iteration matrix can be 0 or not.
- **Consistent ordering scheme** = ordering with which the diagonal blocks are all 0 (C-M, R-B)
- The natural ordering can be a consistent ordering scheme for the **Line (Block) Jacobi** and is NOT a consistent scheme for the pointwise Jacobi **except for 1D problems**

Properties of Jacobi Iteration Matrix

2) Diagonal Dominance

$$-\tilde{D}_m^e h_j^y \bar{\phi}_{me} - \tilde{D}_m^w h_j^y \bar{\phi}_{mw} - \tilde{D}_m^s h_i^x \bar{\phi}_{ms} - \tilde{D}_m^n h_i^x \bar{\phi}_{mn} + (\tilde{D}_m^e h_j^y + \tilde{D}_m^w h_j^y + \tilde{D}_m^s h_i^x + \tilde{D}_m^n h_i^x + \Sigma_r V_m) \bar{\phi}_m = \bar{S} V_m$$

$$s_i^{OD} = \sum_{j \neq i} |a_{ij}| < a_{ii} \quad \forall i \quad \leftarrow \text{Diagonal Dominance}$$

– Jacobi Iteration Matrix T

$$T = [t_{ij}] = D^{-1}(L + U) = \begin{bmatrix} -\frac{a_{ij}}{a_{ii}} \end{bmatrix} \quad \otimes t_{kk} = 0$$

$$\sum |t_{ij}| < 1$$

– Let u be the fundamental eigenvector of T with λ being the corresponding eigenvalue

Namely, $Tu = \lambda u$

Properties of Jacobi Iteration Matrix

2) Diagonal Dominance - conti.

– k -th row $\sum_j t_{kj} u_j = \lambda u_k$

$$\sum_{j \neq k} t_{kj} u_j = \lambda u_k \quad \because t_{kk} = 0$$

$$u = [u_1, \dots, u_k, \dots, u_n]$$

Let u_k be the maximum element (Absolute)

$$|\lambda| |u_k| \leq \sum_{j \neq k} |t_{kj}| |u_j| \leq \sum_{j \neq k} |t_{kj}| |u_k| \quad \because |u_k| \text{ is Max}$$

$$\lambda \leq \sum_{j \neq k} |t_{kj}| < 1$$

♠ Discussion

- If A is diagonally dominant, then $\rho(T) < 1$ where $T = D^{-1}(L + U)$
- As absorption increases, $\rho(T)$ approaches to 0. So converges **faster**.
- As the diffusion coefficient increases, $\rho(T)$ approaches to 1. So converges **slower**.

Properties of Jacobi Iteration Matrix

3) Eigenvalues of T appear in \pm pairs for the consistent ordering scheme

$$T = D^{-1}(L + U) = \begin{bmatrix} 0 & D_1^{-1}U_1 & & & 0 \\ D_2^{-1}L_2 & 0 & D_2^{-1}U_2 & & \\ & D_3^{-1}L_3 & 0 & D_3^{-1}U_3 & \\ & & D_4^{-1}L_4 & 0 & D_4^{-1}U_4 \\ 0 & & & D_5^{-1}L_5 & 0 \end{bmatrix} \quad (\text{Block Tridiagonal Structure})$$

Let $T\psi = \lambda\psi$, then $-\lambda$ is an eigenvalue of T as well

Proof) – Let $\phi_i = (-1)^i \psi_i$

$$T\psi = \lambda\psi$$

$$D^{-1}(L + U)\psi = \lambda\psi$$

$$D_i^{-1}L_i\psi_{i-1} + D_i^{-1}U_i\psi_{i+1} = \lambda\psi_i \quad \rightarrow L_i\psi_{i-1} + U_i\psi_{i+1} = \lambda D_i\psi_i$$

– Apply T to ϕ ($T\phi$)

$$D_i^{-1}L_i\phi_{i-1} + D_i^{-1}U_i\phi_{i+1} = D_i^{-1}L_i(-1)^{i-1}\psi_{i-1} + D_i^{-1}U_i(-1)^{i+1}\psi_{i+1}$$

$$= -(-1)^i [D_i^{-1}L_i\psi_{i-1} + D_i^{-1}U_i\psi_{i+1}] = -(-1)^i \lambda\psi_i = -\lambda\phi_i$$

$\therefore -\lambda$ is another eigenvalue corresponding $\phi = [(-1)^i \psi_i]$

$$\phi = [(-1)^i \psi_i] = \begin{bmatrix} -\psi_1 \\ \psi_2 \\ (-1)^i \psi_i \\ \vdots \\ (-1)^N \psi_N \end{bmatrix}$$

Properties of SOR Iteration Matrix (1/5)

$$T_{SOR} = M^{-1}N = (D - \omega L)^{-1} [(1 - \omega)D + \omega U] \quad \text{For which } \omega, \rho(T) \text{ gets minimized?}$$

- Relation eigenvalues of Jacobi and SOR iteration matrix

T : SOR Iteration Matrix

T_J : Jacobi Iteration Matrix

–Let $T\phi = \gamma\phi$ and $T_J\psi = \lambda\psi$

$$T\phi = \gamma\phi$$

$$(D - \omega L)^{-1} [(1 - \omega)D + \omega U] \phi = \gamma\phi$$

$$(1 - \omega)D\phi + \omega U\phi = \gamma(D - \omega L)\phi$$

$$\phi = \begin{bmatrix} \phi_1 \\ \phi_2 \\ \phi_i \\ \vdots \\ \phi_N \end{bmatrix}, \psi = \begin{bmatrix} \psi_1 \\ \psi_2 \\ \psi_i \\ \vdots \\ \psi_N \end{bmatrix}$$

$$(1 - \omega)D_i\phi_i + \omega U_i\phi_{i+1} = \gamma(D_i\phi_i - \omega L_i\phi_{i-1}) \rightarrow \text{Assume } \phi_i = \alpha^i\psi_i$$

$$(1 - \omega)D_i\alpha^i\psi_i + \omega U_i\alpha^{i+1}\psi_{i+1} = \gamma(D_i\alpha^i\psi_i - \omega L_i\alpha^{i-1}\psi_{i-1})$$

$$(1 - \omega)D_i\psi_i + \omega U_i\alpha\psi_{i+1} = \gamma D_i\psi_i - \omega \frac{\gamma}{\alpha} L_i\psi_{i-1}$$

$$\omega \left(\frac{\gamma}{\alpha} L_i\psi_{i-1} + U_i\alpha\psi_{i+1} \right) = (\gamma + \omega - 1)D_i\psi_i$$

Properties of SOR Iteration Matrix (1/5)

$$T_{SOR} = M^{-1}N = (D - \omega L)^{-1} [(1 - \omega)D + \omega U] \quad \text{For which } \omega, \rho(T) \text{ gets minimized?}$$

- Relation eigenvalues of Jacobi and SOR iteration matrix

– Let $T\phi = \gamma\phi$ and $T_J\psi = \lambda\psi$

$$T\phi = \gamma\phi$$

$$(D - \omega L)^{-1} [(1 - \omega)D + \omega U] \phi = \gamma\phi$$

$$(1 - \omega)D\phi + \omega U\phi = \gamma(D - \omega L)\phi$$

$$(1 - \omega)D_i\phi_i + \omega U_{i,i+1}\phi_{i+1} = \gamma(D_i\phi_i - \omega L_i\phi_{i-1})$$

$$\omega L_i\phi_{i-1} + \omega U_{i,i+1}\phi_{i+1} = (\gamma + \omega - 1)D_i\phi_i$$

$$\omega \left(\frac{\gamma}{\alpha} L_i\psi_{i-1} + U_{i,i+1}\alpha\psi_{i+1} \right) = (\gamma + \omega - 1)D_i\psi_i$$

T : SOR Iteration Matrix

T_J : Jacobi Iteration Matrix

$$\phi = \begin{bmatrix} \phi_1 \\ \phi_2 \\ \phi_i \\ \vdots \\ \phi_N \end{bmatrix}, \psi = \begin{bmatrix} \psi_1 \\ \psi_2 \\ \psi_i \\ \vdots \\ \psi_N \end{bmatrix}$$

Assume $\phi_i = \alpha^i \psi_i$

Properties of SOR Iteration Matrix (2/5)

$$\omega \left(\frac{\gamma}{\alpha} L_i \psi_{i-1} + U_i \alpha \psi_{i+1} \right) = (\gamma + \omega - 1) D_i \psi_i$$

$$T_j \psi = \lambda \psi$$

$$L_i \psi_{i-1} + U_i \psi_{i+1} = \lambda D_i \psi_i$$



$$\omega \left(\frac{\gamma}{\alpha} L_i \psi_{i-1} + U_i \alpha \psi_{i+1} \right) = (\gamma + \omega - 1) \frac{1}{\lambda} (L_i \psi_{i-1} + U_i \psi_{i+1})$$

$$\omega \frac{\gamma}{\alpha} = (\gamma + \omega - 1) \frac{1}{\lambda}$$

$$\rightarrow \frac{\gamma}{\alpha^2} = 1 \rightarrow \alpha = \sqrt{\gamma}$$

$$\omega \alpha = (\gamma + \omega - 1) \frac{1}{\lambda}$$

$$\phi_i = \sqrt{\gamma}^i \psi_i$$

$$\phi = \begin{bmatrix} \sqrt{\gamma} \psi_1 \\ \gamma \psi_2 \\ \sqrt{\gamma}^i \psi_i \\ \vdots \\ \sqrt{\gamma}^N \psi_N \end{bmatrix}$$

$$\sqrt{\gamma} \omega \lambda = (\gamma + \omega - 1)$$

If $\omega = 1$, T is iteration matrix of Gauss seidel

$$\sqrt{\gamma} \lambda = \gamma \rightarrow \gamma = \lambda^2$$

→ Gauss Seidel is two times faster than Jacobi.

Properties of SOR Iteration Matrix (3/5)

$\sqrt{\gamma \omega \lambda} = (\gamma + \omega - 1) \rightarrow \gamma \omega^2 \lambda^2 = (\gamma + \omega - 1)^2 \rightarrow$ Find root γ which is dependent on ω .

$$\gamma^2 + [2(\omega - 1) - \omega^2 \lambda^2] \gamma + (\omega - 1)^2 = 0$$

$$\gamma^2 - 2 \left[(1 - \omega) + \frac{\omega^2 \lambda^2}{2} \right] \gamma + (\omega - 1)^2 = 0$$

$$\gamma = (1 - \omega) + \frac{\omega^2 \lambda^2}{2} \pm \sqrt{\omega^2 \lambda^2 (1 - \omega) + \left(\frac{\omega^2 \lambda^2}{2} \right)^2} = (1 - \omega) + \frac{\omega^2 \lambda^2}{2} \pm \omega |\lambda| \sqrt{(1 - \omega) + \frac{\omega^2 \lambda^2}{4}}$$

$$\begin{cases} \gamma_+ = (1 - \omega) + \frac{\omega^2 \lambda^2}{2} + \omega \lambda \sqrt{\delta} \\ \gamma_- = (1 - \omega) + \frac{\omega^2 \lambda^2}{2} - \omega \lambda \sqrt{\delta} \end{cases}$$

where $\delta = (1 - \omega) + \frac{\omega^2 \lambda^2}{4}$ when will $|\gamma|$ be minimum?

Properties of SOR Iteration Matrix (4/5)

- Proof of Property-conti.

- If $\delta < 0$, then γ is complex.

$$\delta = 1 - \omega + \frac{\omega^2 \lambda^2}{4} < 0$$

$$|\gamma|^2 = \left(1 - \omega + \frac{\omega^2 \lambda^2}{2}\right)^2 - \omega^2 \lambda^2 (-\delta) \rightarrow \gamma = \omega - 1$$

- If $\delta = 0$

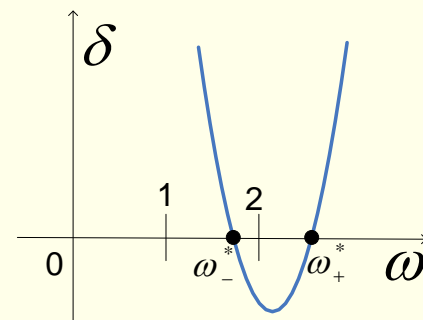
$$\delta = 1 - \omega + \frac{\omega^2 \lambda^2}{4} = 0$$

$$\omega^2 \lambda^2 - 4\omega + 4 = 0$$

$$\omega = \frac{1}{\lambda^2} \left(2 \pm \sqrt{4 - 4\lambda^2}\right) = \frac{2}{\lambda^2} \left(1 \pm \sqrt{1 - \lambda^2}\right)$$

$$\begin{cases} \omega_+^* = \frac{2}{\lambda^2} \left(1 + \sqrt{1 - \lambda^2}\right) = \frac{2}{1 - \sqrt{1 - \lambda^2}} > 2 \\ \omega_-^* = \frac{2}{\lambda^2} \left(1 - \sqrt{1 - \lambda^2}\right) = \frac{2}{1 + \sqrt{1 - \lambda^2}} > 1 \end{cases}$$

$$(\lambda = \rho(T_J) < 1)$$



Properties of SOR Iteration Matrix (5/5)

- Proof of Property-conti.

As $\omega \uparrow \rightarrow \gamma^+ = (1 - \omega) + \frac{\omega^2 \lambda^2}{2} \uparrow \Rightarrow$ Choose smaller ω (ω_-^*)

$$\omega_{opt} = \omega_-^* = \frac{2}{1 + \sqrt{1 - \lambda^2}} = \frac{2}{1 + \sqrt{1 - \rho_{GS}}}$$

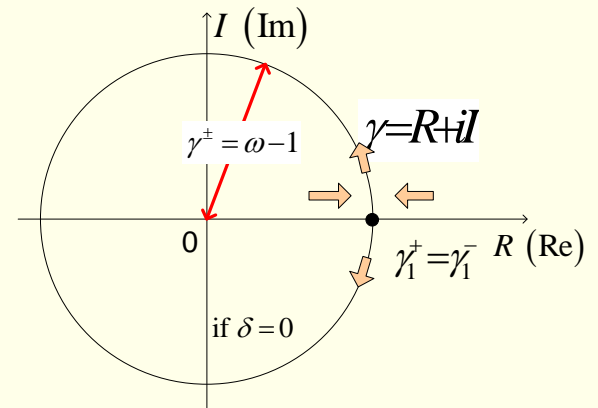
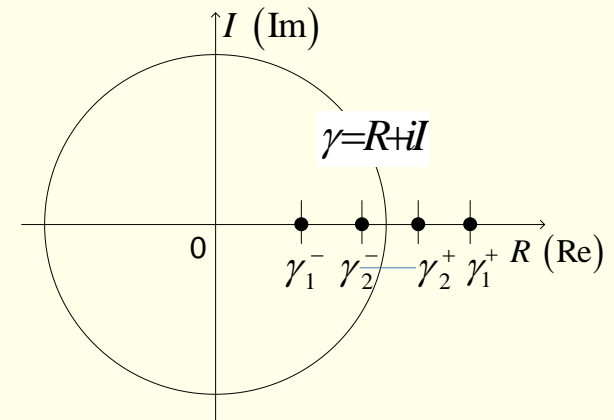
– When $\omega = \omega_{opt}$

$$\delta = 1 - \omega + \frac{\omega^2 \lambda^2}{4} = 0$$

$$\omega^2 \lambda^2 = 4(\omega - 1)$$

$$\begin{aligned} \gamma_{\pm} &= (1 - \omega) + \frac{\omega^2 \lambda^2}{2} \pm \omega \lambda \sqrt{\delta} \\ &= (1 - \omega) + \frac{\omega^2 \lambda^2}{2} \\ &= (1 - \omega) + 2(\omega - 1) \\ &= \omega - 1 \end{aligned}$$

$\therefore \gamma = \omega - 1 > 1$ for ω_+^*



Estimation of ρ_{GS}

- Eigenvalue Problem for Iteration Matrix

$$M^{-1}N\phi = \lambda\phi \quad \text{Inverse Power Method : } M\phi^{(l)} = \frac{1}{\lambda}N\phi^{(l-1)}? \quad M = D - L$$

- Solve an arbitrary linear system iteratively

$$M\phi^{(l)} = \tilde{b} + N\phi^{(l-1)}$$

$$M\phi^{(l-1)} = \tilde{b} + N\phi^{(l-2)}$$

$$M(\phi^{(l)} - \phi^{(l-1)}) = N(\phi^{(l-1)} - \phi^{(l-2)}) \rightarrow \tilde{e}^{(l)} = M^{-1}N\tilde{e}^{(l-1)} = T\tilde{e}^{(l-1)} = \rho\tilde{e}^{(l-1)}$$

$$\rho(M^{-1}N) = \frac{\|\tilde{e}^{(l)}\|}{\|\tilde{e}^{(l-1)}\|}$$

- Practical way to obtain ρ_{GS}
 - perform a few G-S iteration with $b=0$
 - monitor pseudo error reduction, then determine ρ_{GS}

Cyclic-Chebyshev Iterative Method

- Disadvantage of SOR

- Good performance with ω_{opt}
- Bad performance when ω is out of an ω_{opt} , even a little.
- Time and cost are needed for obtaining ω_{opt}

- Normal Iterative Scheme

$$A\phi = b$$

$$(M - N)\phi = b$$

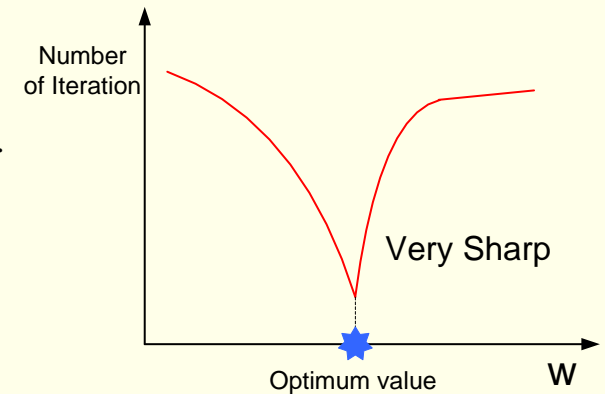
$$M\phi^{(l)} = N\phi^{(l-1)} + b$$

$$\phi^{(l)} = M^{-1}N\phi^{(l-1)} + M^{-1}b$$

$$= T\phi^{(l-1)} + g \quad \text{for Jacobi}$$

$$= D^{-1}(L + U)\phi^{(l-1)} + g$$

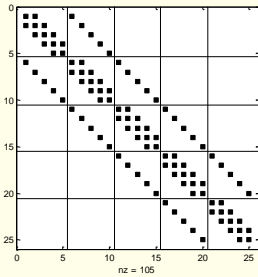
$$= D^{-1}L\phi^{(l-1)} + D^{-1}U\phi^{(l-1)} + g$$



Cyclic-Chebyshev Iterative Method

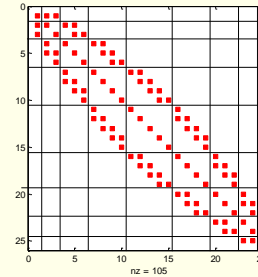
- Solution vector of consistently ordered scheme for Jacobi

[Natural ordering for Line Jacobi]



$$\phi = \begin{bmatrix} \begin{bmatrix} \phi_1 \\ \vdots \\ \phi_m \end{bmatrix} \\ \begin{bmatrix} \phi_{m+1} \\ \vdots \\ \phi_{2m} \end{bmatrix} \\ \begin{bmatrix} \phi_{2m+1} \\ \vdots \\ \phi_{3m} \end{bmatrix} \\ \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{bmatrix} = \begin{bmatrix} \rightarrow \\ \phi_1 \\ \rightarrow \\ \phi_2 \\ \rightarrow \\ \phi_3 \\ \rightarrow \\ \phi_4 \\ \vdots \end{bmatrix}$$

[Cuthill-McKee ordering]



$$\phi = \begin{bmatrix} \begin{bmatrix} \phi_1 \end{bmatrix} \\ \begin{bmatrix} \phi_2 \\ \phi_3 \end{bmatrix} \\ \begin{bmatrix} \phi_4 \\ \phi_5 \\ \phi_6 \end{bmatrix} \\ \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{bmatrix} = \begin{bmatrix} \rightarrow \\ \phi_1 \\ \rightarrow \\ \phi_2 \\ \rightarrow \\ \phi_3 \\ \rightarrow \\ \phi_4 \\ \rightarrow \\ \phi_5 \\ \vdots \end{bmatrix}$$

Cyclic-Chebyshev Iterative Method

- Change the sequence of variables

$$\begin{bmatrix} \vec{\phi}_1 \\ \vec{\phi}_2 \\ \vec{\phi}_3 \\ \vec{\phi}_4 \\ \vec{\phi}_5 \\ \vec{\phi}_6 \\ \vec{\phi}_7 \end{bmatrix} = \begin{bmatrix} 0 & D_1^{-1}U_1 & & & & & \\ D_2^{-1}L_2 & 0 & D_2^{-1}U_2 & & & & \\ & D_3^{-1}L_3 & 0 & D_3^{-1}U_3 & & & \\ & & D_4^{-1}L_4 & 0 & D_4^{-1}U_4 & & \\ & & & D_5^{-1}L_5 & 0 & D_5^{-1}U_5 & \\ & & & & D_6^{-1}L_6 & 0 & D_6^{-1}U_6 \\ & & & & & D_7^{-1}L_7 & 0 \end{bmatrix} \begin{bmatrix} \vec{\phi}_1 \\ \vec{\phi}_2 \\ \vec{\phi}_3 \\ \vec{\phi}_4 \\ \vec{\phi}_5 \\ \vec{\phi}_6 \\ \vec{\phi}_7 \end{bmatrix} + \vec{g}$$

$$\begin{cases} \vec{\phi}_1 = D_1^{-1}U_1\vec{\phi}_2 + \vec{g}_1 \\ \vec{\phi}_3 = D_3^{-1}L_3\vec{\phi}_2 + D_3^{-1}U_3\vec{\phi}_4 + \vec{g}_3 \\ \vec{\phi}_5 = D_5^{-1}L_5\vec{\phi}_4 + D_5^{-1}U_5\vec{\phi}_6 + \vec{g}_5 \\ \vdots \end{cases}$$

$$\begin{aligned} \phi^{(l)} &= T\phi^{(l-1)} + \vec{g} \\ \vec{\phi}_i^{(l)} &= D_i^{-1}L_i\vec{\phi}_{i-1}^{(l-1)} + D_i^{-1}U_i\vec{\phi}_{i+1}^{(l-1)} + \vec{g} \end{aligned}$$

$$\begin{bmatrix} \vec{\phi}_1 \\ \vec{\phi}_3 \\ \vec{\phi}_5 \\ \vec{\phi}_7 \\ \vec{\phi}_2 \\ \vec{\phi}_4 \\ \vec{\phi}_6 \end{bmatrix} = \begin{bmatrix} D_1^{-1}U_1 & & & & & & \\ D_3^{-1}L_3 & D_3^{-1}U_3 & & & & & \\ & D_5^{-1}L_5 & D_5^{-1}U_5 & & & & \\ & & D_7^{-1}L_7 & 0 & & & \\ & & & D_2^{-1}L_2 & D_2^{-1}U_2 & & \\ & & & & D_4^{-1}L_4 & D_4^{-1}U_4 & \\ & & & & & D_6^{-1}L_6 & D_6^{-1}U_6 \end{bmatrix} \begin{bmatrix} \vec{\phi}_2 \\ \vec{\phi}_4 \\ \vec{\phi}_6 \\ \vec{\phi}_1 \\ \vec{\phi}_3 \\ \vec{\phi}_5 \\ \vec{\phi}_6 \end{bmatrix} + \vec{g}'$$

$$\begin{bmatrix} \vec{\phi}_1 \\ \vec{\phi}_2 \end{bmatrix} = \begin{bmatrix} T_1 & 0 \\ 0 & T_2 \end{bmatrix} \begin{bmatrix} \vec{\phi}_2 \\ \vec{\phi}_1 \end{bmatrix} + \begin{bmatrix} \vec{g}_1 \\ \vec{g}_2 \end{bmatrix}$$

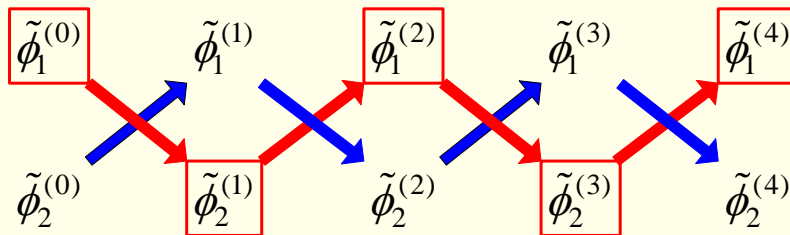
Cyclic-Chebyshev Iterative Method

- Change the order of variables

$$\begin{bmatrix} \tilde{\phi}_1 \\ \tilde{\phi}_2 \end{bmatrix} = \begin{bmatrix} T_1 & 0 \\ 0 & T_2 \end{bmatrix} \begin{bmatrix} \tilde{\phi}_2 \\ \tilde{\phi}_1 \end{bmatrix} + \begin{bmatrix} g_1 \\ g_2 \end{bmatrix}$$

$$\Rightarrow \begin{cases} \tilde{\phi}_1^{(l)} = T_1 \tilde{\phi}_2^{(l-1)} + g_1 \\ \tilde{\phi}_2^{(l)} = T_2 \tilde{\phi}_1^{(l-1)} + g_2 \end{cases}$$

- Cyclic iterative method (advantage of consistently ordering)



- One path is enough.

Cyclic-Chebyshev Iterative Method

- Extrapolation

- Jacobi

$$\phi_{Jacobi}^{(l)} = T \phi^{(l-1)} + g$$

- Extrapolation

$$\phi^{(l)} = \alpha^{(l)} \phi_{Jacobi}^{(l)} + (1 - \alpha^{(l)} + \beta^{(l)}) \phi^{(l-1)} - \beta^{(l)} \phi^{(l-2)}$$

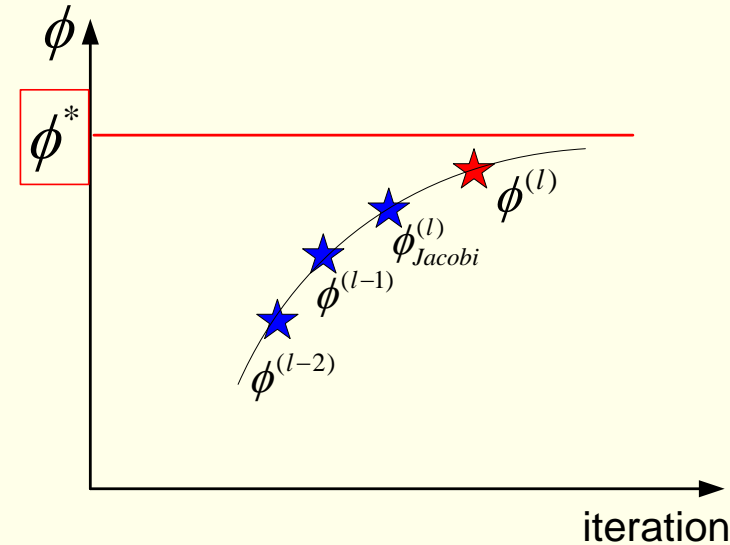
- l -th error

$$\left. \begin{aligned} \phi^{(l)} &= \alpha^{(l)} (T \phi^{(l-1)} + g) + (1 - \alpha^{(l)} + \beta^{(l)}) \phi^{(l-1)} - \beta^{(l)} \phi^{(l-2)} \\ \phi^* &= \alpha^{(l)} (T \phi^* + g) + (1 - \alpha^{(l)} + \beta^{(l)}) \phi^* - \beta^{(l)} \phi^* \end{aligned} \right\} (\phi^* : \text{Exact solution})$$

$$e^{(l)} = \alpha^{(l)} T e^{(l-1)} + (1 - \alpha^{(l)} + \beta^{(l)}) e^{(l-1)} - \beta^{(l)} e^{(l-2)}$$

$$e^{(l)} = \sum_{i=1}^n p^{(l)}(\lambda_i) c_i u_i \quad \text{with } e^{(0)} = \sum_{i=1}^n c_i u_i$$

- Should minimize the l -th error \rightarrow Chebyshev polynomial



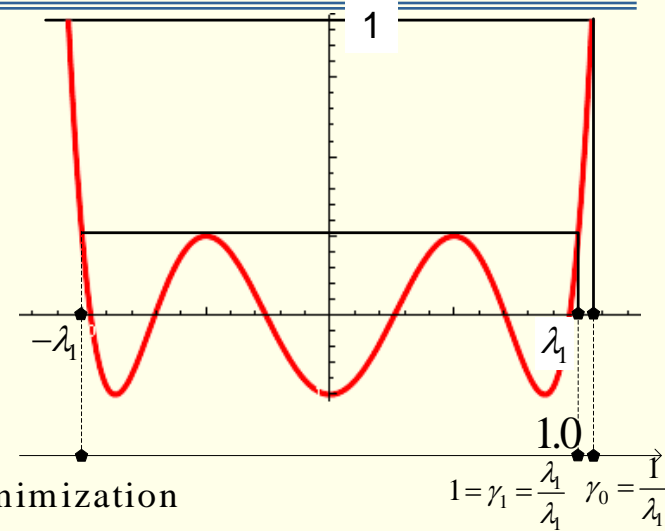
Cyclic-Chebyshev Iterative Method

– Change of variable

$$\left(\begin{array}{l} \gamma_i = \frac{\lambda_i}{\lambda_1} , \quad \gamma_0 = \frac{1}{\lambda_1} \end{array} \right)$$

$$e^{(l)} = \sum_{i=1}^n p^{(l)}(\lambda_i) c_i u_i = \sum_{i=1}^n \eta^{(l)}(\gamma_i) c_i u_i$$

$$= \eta^{(l)}(\gamma_0) \sum_{i=1}^n \frac{\eta^{(l)}(\gamma_i)}{\eta^{(l)}(\gamma_0)} c_i u_i \quad \rightarrow \quad \eta^{(l)}(\gamma) = \frac{T_l(\gamma)}{T_l(\gamma_1)} = \tilde{T}_l(\gamma) \text{ for minimization}$$



• Recurrence relation of Chebyshev polynomial

$$T_l(\gamma) = 2\gamma T_{l-1}(\gamma) - T_{l-2}(\gamma)$$

$$\frac{T_l(\gamma)}{T_l(\gamma_0)} = 2\gamma \frac{T_{l-1}(\gamma)}{T_l(\gamma_0)} - \frac{T_{l-2}(\gamma)}{T_l(\gamma_0)} = 2\gamma \frac{T_{l-1}(\gamma_0)}{T_l(\gamma_0)} \frac{T_{l-1}(\gamma)}{T_{l-1}(\gamma_0)} - \frac{T_{l-2}(\gamma_0)}{T_l(\gamma_0)} \frac{T_{l-2}(\gamma)}{T_{l-2}(\gamma_0)}$$

$$\tilde{T}_l(\gamma) = 2\gamma \frac{T_{l-1}(\gamma_0)}{T_l(\gamma_0)} \tilde{T}_{l-1}(\gamma) - \frac{T_{l-2}(\gamma_0)}{T_l(\gamma_0)} \tilde{T}_{l-2}(\gamma)$$

Cyclic-Chebyshev Iterative Method

- Extrapolation parameters

– Compare two formula

$$\begin{cases} \tilde{T}_l(\gamma) = 2\gamma \frac{T_{l-1}(\gamma_0)}{T_l(\gamma_0)} \tilde{T}_{l-1}(\gamma) - \frac{T_{l-2}(\gamma_0)}{T_l(\gamma_0)} \tilde{T}_{l-2}(\gamma) \\ \eta^{(l)}(\gamma) = \alpha^{(l)} \lambda_1 \gamma \eta^{(l-1)}(\gamma) + (1 - \alpha^{(l)} + \beta^{(l)}) \eta^{(l-1)}(\gamma) - \beta^{(l)} \eta^{(l-2)}(\gamma) \end{cases}$$

$$\Rightarrow \alpha^{(l)} \lambda_1 = 2 \frac{T_{l-1}(\gamma_0)}{T_l(\gamma_0)} \quad \alpha^{(l)} = \frac{2}{\lambda_1} \frac{T_{l-1}(\gamma_0)}{T_l(\gamma_0)} = \frac{2}{\lambda_1} \frac{T_{l-1}\left(\frac{1}{\lambda_1}\right)}{T_l\left(\frac{1}{\lambda_1}\right)} = 2\gamma_0 \frac{T_{l-1}(\gamma_0)}{T_l(\gamma_0)}$$

$$\beta^{(l)} = \frac{T_{l-2}(\gamma_0)}{T_l(\gamma_0)} = \frac{T_{l-2}\left(\frac{1}{\lambda_1}\right)}{T_l\left(\frac{1}{\lambda_1}\right)}$$

-Aftermath

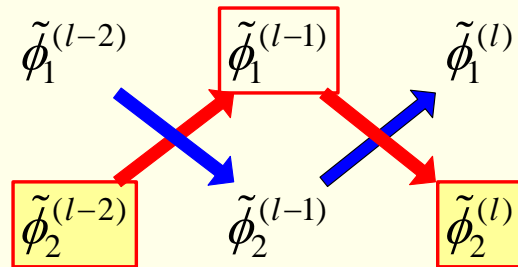
$$1 - \alpha^{(l)} + \beta^{(l)} = 1 - 2\gamma_0 \frac{T_{l-1}(\gamma_0)}{T_l(\gamma_0)} + \frac{T_{l-2}(\gamma_0)}{T_l(\gamma_0)} = 1 - \frac{T_l(\gamma_0) + T_{l-2}(\gamma_0)}{T_l(\gamma_0)} + \frac{T_{l-2}(\gamma_0)}{T_l(\gamma_0)} = 0$$

Cyclic-Chebyshev Iterative Method

- Advantage of cyclic iterative method for applying Chebyshev

$$\phi^{(l)} = \alpha^{(l)} \phi_{Jacobi}^{(l)} + (1 - \alpha^{(l)} + \beta^{(l)}) \phi^{(l-1)} - \beta^{(l)} \phi^{(l-2)}$$

$$\Rightarrow \phi^{(l)} = \alpha^{(l)} \phi_{Jacobi}^{(l)} - \beta^{(l)} \phi^{(l-2)} \quad (\because 1 - \alpha^{(l)} + \beta^{(l)} = 0)$$



– Do it only for half.

– Convergence ratio $\rho_{CCSi} = \frac{1}{2}(\omega_{opt} - 1)$

– Total calculating quantity

$$\left[\text{calculate half (each step)} \right] \times \left[\text{calculate double (iteration)} \right] = 1$$

– Good for not exact ω_{opt}

Consistently Ordered Matrix

Let the $N \times N$ matrix A be partitioned into the form

$$A = \begin{bmatrix} A_{1,1} & \cdots & A_{1,q} \\ \vdots & & \vdots \\ A_{q,1} & \cdots & A_{q,q} \end{bmatrix}, \quad (9-2.1)$$

where $A_{i,j}$ is an $n_i \times n_j$ submatrix and $n_1 + \cdots + n_q = N$.

Definition 9-2.1 The $q \times q$ block matrix A of (9-2.1) is said to have *Property \mathcal{A}* if there exists two disjoint nonempty subsets S_R and S_B of $\{1, 2, \dots, q\}$ such that $S_R \cup S_B = \{1, 2, \dots, q\}$ and such that if $A_{i,j} \neq 0$ and $i \neq j$, then $i \in S_R$ and $j \in S_B$ or $i \in S_B$ and $j \in S_R$.

Definition 9-2.2. The $q \times q$ block matrix A of (9-2.1) is said to be *consistently ordered* if for some t there exist disjoint nonempty subsets S_1, \dots, S_t of $\{1, 2, \dots, q\}$ such that $\bigcup_{i=1}^t S_i = \{1, \dots, q\}$ and such that if $A_{i,j} \neq 0$ with $i \neq j$ and S_k is the subset containing i , then $j \in S_{k+1}$ if $j > i$ and $j \in S_{k-1}$ if $j < i$.