

Lecture 14: Recall:

Splitting choice 3: Successive overrelaxation method (SOR)

Consider the iterative scheme =

(Suppose $A = L + D + U$)

$$L \vec{x}^{k+1} + D \vec{y}^{k+1} + U \vec{x}^k = \vec{b} \quad (*)$$

$$\vec{x}^{k+1} = \vec{x}^k + \omega (\vec{y}^{k+1} - \vec{x}^k) \quad (**)$$

$$\Leftrightarrow \vec{y}^{k+1} = \frac{1}{\omega} (\overset{\text{IR}}{\vec{x}^{k+1}} + (\omega - 1) \vec{x}^k)$$

Putting (**) into (*):

$$(L + \frac{1}{\omega} D) \vec{x}^{k+1} + \frac{1}{\omega} (\omega U + (\omega - 1) D) \vec{x}^k = \vec{b}$$

$$\text{or } \underbrace{(L + \frac{1}{\omega} D)}_N \vec{x}^{k+1} = \underbrace{(\frac{1}{\omega} D - (D + U))}_P \vec{x}^k + \vec{b}$$

Remark: SOR is equivalent to:

$$A = N - P = \underbrace{\left(L + \frac{1}{\omega}D\right)}_N - \underbrace{\left(\frac{1}{\omega}D - (D+U)\right)}_P$$

Or equivalently, $A = (a_{ij})$

$$\left\{ \begin{array}{l} a_{11} y_1^{k+1} + a_{12} x_2^k + \dots + a_{1n} x_n^k = b_1 \text{ for } x_1^{k+1} = x_1^k + \omega(y_1^{k+1} - x_1^k) \\ a_{21} x_1^{k+1} + a_{22} y_2^{k+1} + a_{23} x_3^k + \dots + a_{2n} x_n^k = b_2 \text{ for } x_2^{k+1} = x_2^k + \omega(y_2^{k+1} - x_2^k) \\ \vdots \\ a_{n1} x_1^{k+1} + a_{n2} x_2^{k+1} + \dots + a_{nn} y_n^{k+1} = b_n \text{ for } x_n^{k+1} = x_n^k + \omega(y_n^{k+1} - x_n^k) \end{array} \right.$$

• SOR = Gauss-Seidel if $\omega = 1$.

Condition for convergence

Theorem: The necessary condition (not sufficient) for SOR to converge is $0 < \omega < 2$.

Theorem: If A is strictly diagonally dominant (SDD), then SOR converges if $0 < \omega \leq 1$.

Proof: We need to show $\rho(M_{\text{SOR}}) < 1$ if $0 < \omega \leq 1$.

We'll show it by contradiction.

Suppose $\exists$ eigenvalue λ such that $|\lambda| \geq 1$. Then:

$$\det(\lambda I - M_{\text{SOR}}) = 0$$

$$\therefore \det(\lambda I - (D + \omega L)^{-1}((1 - \omega)D - \omega U)) = 0$$

$$\Rightarrow \det\left(\underbrace{\lambda}_{\neq 0} \underbrace{(D + \omega L)^{-1}}_D \left(\underbrace{(D + \omega L)}_C - \frac{1}{\lambda}((1 - \omega)D - \omega U) \right) \right) = 0$$

$$\Rightarrow \det(C) = 0 \quad (\because \lambda \neq 0, (D + \omega L) \text{ is invertible})$$

Note: $\omega(1 - \frac{1}{|\lambda|}) \leq (1 - \frac{1}{|\lambda|}) \Rightarrow (1 - \frac{1}{|\lambda|}(1-\omega)) \geq \omega$

Now,

$$|C_{ii}| = \left| 1 - \frac{1}{\lambda}(1-\omega) \right| |a_{ii}| \geq \left[1 - \frac{1}{|\lambda|}(1-\omega) \right] |a_{ii}|$$

$$\geq \omega |a_{ii}| > \omega \sum_{\substack{j=1 \\ j \neq i}}^n |a_{ij}| \geq \underbrace{\omega \sum_{j=1}^{i-1} |a_{ij}| + \frac{\omega}{|\lambda|} \sum_{j=i+1}^n |a_{ij}|}_{\sum_{\substack{j=1 \\ j \neq i}}^n |C_{ij}|}$$

$$|C_{ii}| > \sum_{\substack{j=1 \\ j \neq i}}^n |C_{ij}| \text{ for all } i$$

$$\sum_{\substack{j=1 \\ j \neq i}}^n |C_{ij}|$$

$\therefore C$ is SDD $\Rightarrow C$ is non-singular!!

$$\Rightarrow \det(C) \neq 0$$

Contradiction.

$$\left((D + \omega L) - \frac{1}{\lambda} \underbrace{\left((1-\omega)D - \omega U \right)}_C \right)$$

Optimal parameter ω_{opt} for SOR method

Definition: Consider the system $A\vec{x} = \vec{b}$. Let $A = \overset{\text{lower}}{\downarrow} L + \overset{\text{diagonal}}{\downarrow} D + \overset{\text{upper}}{\nwarrow} U$.

If the eigenvalues of: $\alpha D^{-1}L + \frac{1}{\alpha} D^{-1}U$ ($\alpha \neq 0$) are independent of α . Then, the matrix A is said to be consistently ordered.

Example of consistently ordered matrices

1. Tridiagonal matrix: $\begin{pmatrix} \lambda_1 & * & & 0 \\ * & \lambda_2 & & \\ & & \ddots & * \\ 0 & & * & \lambda_n \end{pmatrix}$

2. Block tridiagonal matrix:

$$\begin{pmatrix} \boxed{D_1} & \boxed{T_{12}} & & 0 \\ \boxed{T_{21}} & \boxed{D_2} & \boxed{T_{23}} & \\ & & \ddots & \\ 0 & & & \boxed{D_p} \end{pmatrix}$$

where $D_i =$ diagonal matrix

Theorem: [D. Young] Assume:

1. $0 < \omega < 2$
2. $M_J = N_J^{-1} P_J$ has only real eigenvalues
3. $\beta \stackrel{\text{def}}{=} \rho(M_J) < 1$
4. A is consistently ordered.

Then: $\rho(M_{SOR}) < 1$

Also, the optimal parameter ω_{opt} for fastest convergence

$$\omega_{opt} = \frac{2}{1 + \sqrt{1 - \beta^2}} \quad \text{and}$$

$$\rho(M_{SOR, \omega_{opt}}) = \omega_{opt} - 1$$

Example: $A\vec{x} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 12 \\ 21 \end{pmatrix}$

Then: $\alpha D^{-1}L + \frac{1}{\alpha} D^{-1}U = \begin{pmatrix} 0 & -\frac{1}{10}\alpha \\ \frac{-\alpha}{10} & 0 \end{pmatrix}$

$\therefore$ Char. poly = $\lambda^2 - \left(\frac{-1}{10\cancel{\alpha}}\right)\left(\frac{-\cancel{\alpha}}{10}\right) = 0 \Rightarrow \lambda^2 - \frac{1}{100} = 0$

$\therefore$ A is consistently ordered.

(independent of α)

Also, M_J has only real eigenvalues and $\lambda_1 = \frac{1}{10}$, $\lambda_2 = -\frac{1}{10}$.

$\therefore$ SOR converges if $0 < \omega < 2$.

$\therefore \rho(M_J) < 1$

$\therefore$ Optimal $\omega_{\text{opt}} = \frac{2}{1 + \sqrt{1 - \rho(M_J)^2}} = \frac{2}{1 + \sqrt{1 - \left(\frac{1}{10}\right)^2}} = 1.0025126$

$\rho(M_{\text{SOR}, \omega_{\text{opt}}}) = \omega_{\text{opt}} - 1 = 0.0025126.$