

221900180 田永铭 计算方法作业 12

第7章 18. 解: $\|A\|_1 = 1.1$, $\|A\|_2 = 0.8$, $\|A\|_F = 0.825$, $\|A\|_\infty = 0.8426$.

这题上次作业已做过, 这里直接给答案!

31. 解:

证明:

(1) $\because A$ 对称正定, $\therefore A^T = A, A \succ 0$. 又有 $A = LDL^T = W^T W$, 其中 $W = D^{\frac{1}{2}} L^T$.

$$\therefore \text{cond}(A)_2 = \frac{|\lambda_{\max}(A)|}{|\lambda_{\min}(A)|} = \sqrt{\frac{\lambda_{\max}(A^T A)}{\lambda_{\min}(A^T A)}} = \sqrt{\frac{\lambda_{\max}(W^T W)^2}{\lambda_{\min}(W^T W)^2}} = \sqrt{\left(\frac{\lambda_{\max}(W^T W)}{\lambda_{\min}(W^T W)}\right)^2}$$

$$= [\text{cond}(W)_2]^2. \quad \square.$$

(2) 由 (1): $\text{cond}(A)_2 = [\text{cond}(W)_2]^2$, \therefore 欲证 $\text{cond}(A_2) = \text{cond}(W^T)_2 \text{cond}(W)_2$,
只需证: $\text{cond}(W)_2 = \text{cond}(W^T)_2$.

$$\text{而 } \text{cond}(W^T)_2 = \sqrt{\frac{\lambda_{\max}(W W^T)}{\lambda_{\min}(W W^T)}} = \sqrt{\frac{\lambda_{\max}(W^T W)}{\lambda_{\min}(W^T W)}} = \text{cond}(W)_2.$$

 \therefore 得证. \square .32. 解: ① $A = \begin{pmatrix} 100 & 99 \\ 99 & 98 \end{pmatrix}$, $A^{-1} = \begin{pmatrix} -98 & 99 \\ 99 & -100 \end{pmatrix}$.

$$\therefore \|A\|_\infty = 199, \|A^{-1}\|_\infty = 199.$$

$$\therefore \text{cond}(A)_\infty = \|A\|_\infty \|A^{-1}\|_\infty = 199^2 = 39601.$$

② ~~2. 证 $\text{cond}(A)_2$:~~

$$A^T A = \begin{pmatrix} 100 & 99 \\ 99 & 98 \end{pmatrix} \begin{pmatrix} 100 & 99 \\ 99 & 98 \end{pmatrix} = \begin{pmatrix} 19801 & 19602 \\ 19602 & 19405 \end{pmatrix}.$$

$$\det(\lambda E - A^T A) = \begin{vmatrix} \lambda - 19801 & -19602 \\ -19602 & \lambda - 19405 \end{vmatrix} = 0, \text{ 则 } \lambda_1 = 39205.99997, \lambda_2 = 0.000055006.$$

$$\because A \text{ 为对称阵}, \therefore \text{cond}(A)_2 = \frac{|\lambda|_{\max}}{|\lambda|_{\min}}$$

$$\text{令 } |\lambda E - A| = \begin{vmatrix} \lambda - 100 & -99 \\ -99 & \lambda - 98 \end{vmatrix} = (\lambda - 100)(\lambda - 98) - 99^2 = 0, \therefore \lambda_1 = 198.0050, \lambda_2 = -5.05038 \times 10^{-3}.$$

$$\therefore \text{cond}(A)_2 = \frac{198.0050}{5.05038 \times 10^{-3}} = 39205.96 \approx 39206.$$

34. 证明:

$$\text{cond}(B) = \|B\| \|B^{-1}\|$$

$$(\text{由性质}) \leq \|B^{-1}\| \|B\| \|B\| \|B^{-1}\| = (\|B^{-1}\| \|B\|) (\|B\| \|B^{-1}\|) = \text{cond}(B) \text{cond}(B).$$

证毕!

习题第8章

1. 解: (1) $\because 5 > 2+1, 4 > 1+2, 10 > 2+3$, \therefore 矩阵强对角占优, 由定理知 Jacobi 法 Gauss-Seidel 法均收敛.

$$(2) \textcircled{1} \text{Jacobi: } \begin{cases} x_1^{(k+1)} = -\frac{2}{5}x_2^{(k)} - \frac{1}{5}x_3^{(k)} - \frac{12}{5}, \\ x_2^{(k+1)} = \frac{1}{4}x_1^{(k)} - \frac{1}{2}x_3^{(k)} + 5, \\ x_3^{(k+1)} = \frac{1}{5}x_1^{(k)} - \frac{2}{10}x_2^{(k)} + \frac{3}{10}. \end{cases} \quad \text{取 } X^{(0)} = (0, 0, 0)^T.$$

利用计算机求解, 得: $X^{(18)} = (-3.9999964, 2.9999739, 1.9999999)^T$,
此时迭代18次, $\|X^{(18)} - X^{(17)}\|_{\infty} \approx 0.4 \times 10^{-4} < 10^{-4}$.

$$\textcircled{2} \text{G-S: } \begin{cases} x_1^{(k+1)} = -\frac{2}{5}x_2^{(k)} - \frac{1}{5}x_3^{(k)} - \frac{12}{5}, \\ x_2^{(k+1)} = \frac{1}{4}x_1^{(k+1)} - \frac{1}{2}x_3^{(k)} + 5, \\ x_3^{(k+1)} = \frac{1}{5}x_1^{(k+1)} - \frac{2}{10}x_2^{(k+1)} + \frac{3}{10}, \end{cases} \quad \text{取 } X^{(0)} = (0, 0, 0)^T.$$

利用计算机求解, 得: $X^{(8)} = (-4.000036, 2.999985, 2.000003)^T$,
此时迭代8次, $\|X^{(8)} - X^{(7)}\|_{\infty} \approx 0.9 \times 10^{-4} < 10^{-4}$.

$$5. (1) B = D^{-1}(L+U) = \begin{bmatrix} 0 & -0.4 & -0.6 \\ -0.4 & 0 & -0.8 \\ -0.6 & 0.8 & 0 \end{bmatrix}.$$

$$|\lambda E - B| = (\lambda - 0.8)(\lambda^2 + 0.8\lambda - 0.32). \quad \rho(B) = \lambda_{\max} \approx 1.0928 > 1, \therefore \text{不收敛}.$$

$$\textcircled{2} \text{G-S: } G = (D-L)^{-1}U = \begin{bmatrix} 0 & -0.4 & -0.4 \\ 0 & 0.16 & -0.64 \\ 0 & 0.032 & 0.672 \end{bmatrix}, \quad \|G\|_{\infty} = 0.8.$$

$$\therefore \rho(G) \leq \|G\|_{\infty} = 0.8 < 1, \text{由定理知: G-S 收敛}.$$

$$\textcircled{1} \text{Jacobi: } (2) B = D^{-1}(L+U) = \begin{bmatrix} 0 & -2 & 2 \\ -1 & 0 & 1 \\ -2 & -2 & 0 \end{bmatrix}, \quad |\lambda E - B| = \lambda^3, \quad \rho(B) = 0 < 1, \therefore \text{收敛}.$$

$$\textcircled{2} \text{G-S: } G = (D-L)^{-1}U = \begin{bmatrix} 0 & -2 & 2 \\ 0 & 2 & -3 \\ 0 & 0 & 2 \end{bmatrix}, \quad |\lambda E - G| = \lambda(\lambda-2)^2, \quad \rho(G) = 2 > 1, \therefore \text{不收敛}.$$

$$8. (1) B = D^{-1}(L+U) = \begin{pmatrix} 0 & 0 & \frac{1}{4} & \frac{1}{4} \\ 0 & 0 & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & 0 & 0 \\ \frac{1}{4} & \frac{1}{4} & 0 & 0 \end{pmatrix}, \quad |\lambda E - B| = \lambda^2(\lambda - 0.25)^2, \quad \rho(B) = \lambda_{\max} = 0.25.$$

$$(2) G = (D-L)^{-1}U = \begin{pmatrix} 0 & 0 & \frac{1}{4} & \frac{1}{4} \\ 0 & 0 & \frac{1}{4} & \frac{1}{4} \\ 0 & 0 & \frac{1}{8} & \frac{1}{8} \\ 0 & 0 & \frac{1}{8} & \frac{1}{8} \end{pmatrix}, \quad |\lambda E - G| = \lambda^3(\lambda - 0.25), \quad \rho(G) = \lambda_{\max} = 0.25.$$

$$(3) \because \rho(B) = 0.5 < 1, \rho(G) = 0.25 < 1, 0.25 < 0.5,$$

\therefore Jacobi 法与 Gauss-Seidel 法均收敛, 且 G-S 法收敛更快一点.

