

作业4

李子龙 123033910195

2023年11月15日

1. 解

$$A = \begin{pmatrix} 5 & 2 & 1 \\ -1 & 4 & 2 \\ 2 & -3 & 10 \end{pmatrix} = \begin{pmatrix} 5 & 4 & 10 \\ 4 & 10 \end{pmatrix} - \begin{pmatrix} 1 & 1 & 10 \\ -2 & 3 & 10 \end{pmatrix} - \begin{pmatrix} -2 & -1 & 10 \\ -2 & 3 & 10 \end{pmatrix} = \mathbf{D} - \mathbf{L} - \mathbf{U}$$

$$\mathbf{b} = \begin{pmatrix} -12 & 1 & 10 \\ 20 & 3 & 10 \end{pmatrix}$$

- (1) 由于 A 是严格占优矩阵,所以 Jacobi 迭代法和 Guass–Seidel 迭代法解这个方程组都是收敛的。
- (2) Jacobi 迭代法 对于 Jacobi 迭代法, 其迭代矩阵

$$\mathbf{B} = \mathbf{D}^{-1}(\mathbf{L} + \mathbf{U}) = \begin{pmatrix} -\frac{2}{5} & -\frac{1}{5} \\ \frac{1}{4} & -\frac{1}{2} \\ -\frac{1}{5} & \frac{3}{10} \end{pmatrix} \qquad \mathbf{f} = \mathbf{D}^{-1}\mathbf{b} = \begin{pmatrix} -\frac{12}{5} \\ 5 \\ \frac{3}{10} \end{pmatrix}$$

取迭代初值 $\mathbf{x}^{(0)} = (0,0,0)^{\mathsf{T}}$,记 $\boldsymbol{\epsilon}^{(k)} = \mathbf{x}^{(k)} - \mathbf{x}^{(k-1)}$,有



总共需要18次满足要求。

Guass-Seidel 迭代法 对于 Guass-Seidel 迭代法,

$$G = (D - L)^{-1}U = \begin{pmatrix} 5 & & \\ -1 & 4 & \\ 2 & -3 & 10 \end{pmatrix}^{-1} \begin{pmatrix} -2 & -1 \\ & -2 \end{pmatrix} = \begin{pmatrix} 0 & -0.4 & -0.2 \\ 0 & -0.1 & -0.55 \\ 0 & 0.05 & -0.125 \end{pmatrix}$$
$$f = (D - L)^{-1}b = \begin{pmatrix} 5 & & \\ -1 & 4 & \\ 2 & -3 & 10 \end{pmatrix}^{-1} \begin{pmatrix} -12 \\ 20 \\ 3 \end{pmatrix} = \begin{pmatrix} -2.4 \\ 4.4 \\ 2.1 \end{pmatrix}$$

取迭代初值 $x^{(0)} = (0,0,0)^{\mathsf{T}}$,记 $\epsilon^{(k)} = x^{(k)} - x^{(k-1)}$,有

$$x^{(1)} = Gx^{(0)} + f = (-2.4, 4.4, 2.1)^{\top} \qquad ||\epsilon^{(1)}||_{\infty} = 4.4$$

$$x^{(2)} = Gx^{(1)} + f = (-4.58, 2.805, 2.0575)^{\top} \qquad ||\epsilon^{(2)}||_{\infty} = 2.18$$

$$x^{(3)} = Gx^{(2)} + f = (-3.9335, 2.987875, 1.9830625)^{\top} \qquad ||\epsilon^{(3)}||_{\infty} = 0.6465$$

$$x^{(4)} = Gx^{(3)} + f = (-3.9917625, 3.01052813, 2.00151094)^{\top} \qquad ||\epsilon^{(4)}||_{\infty} = 0.0582625$$

$$x^{(5)} = Gx^{(4)} + f = (-4.00451344, 2.99811617, 2.00033754)^{\top} \qquad ||\epsilon^{(5)}||_{\infty} = 0.01275094$$

$$x^{(6)} = Gx^{(5)} + f = (-3.99931398, 3.00000274, 1.99986362)^{\top} \qquad ||\epsilon^{(6)}||_{\infty} = 0.00519946$$

$$x^{(7)} = Gx^{(6)} + f = (-3.99997382, 3.00007474, 2.00001718)^{\top} \qquad ||\epsilon^{(7)}||_{\infty} = 0.00065984$$

$$x^{(8)} = Gx^{(7)} + f = (-4.00003333, 2.99998307, 2.00000159)^{\top} \qquad ||\epsilon^{(8)}||_{\infty} = 9.16628308 \times 10^{-5}$$

总共需要8次满足要求。

8. 解

$$\mathbf{A} = \begin{pmatrix}
1 & 0 & -\frac{1}{4} & -\frac{1}{4} \\
0 & 1 & -\frac{1}{4} & -\frac{1}{4} \\
-\frac{1}{4} & -\frac{1}{4} & 1 & 0 \\
-\frac{1}{4} & -\frac{1}{4} & 0 & 1
\end{pmatrix} = \begin{pmatrix}
1 & & & \\
& 1 & & \\
& & 1 & \\
& & & 1
\end{pmatrix} - \begin{pmatrix}
& & \frac{1}{4} & \frac{1}{4} \\
& & \frac{1}{4} & \frac{1}{4} \\
& & \frac{1}{4} & \frac{1}{4}
\end{pmatrix} = \mathbf{D} - \mathbf{L} - \mathbf{U}$$

$$\mathbf{b} = \begin{pmatrix}
\frac{1}{2} \\
\frac{1}{2} \\
\frac{1}{2} \\
\frac{1}{2}
\end{pmatrix}$$



(1) 对于 Jacobi 迭代法,

$$\boldsymbol{B}_{0} = \boldsymbol{D}^{-1}(\boldsymbol{L} + \boldsymbol{U}) = \begin{pmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{pmatrix}^{-1} \begin{pmatrix} & & \frac{1}{4} & \frac{1}{4} \\ & & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & & \\ \frac{1}{4} & \frac{1}{4} & & \end{pmatrix} = \begin{pmatrix} & & \frac{1}{4} & \frac{1}{4} \\ & & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & & \\ \frac{1}{4} & \frac{1}{4} & & \end{pmatrix}$$

求它的特征值

$$0 = |\lambda \mathbf{E} - \mathbf{B}_0| = \begin{vmatrix} \lambda & -\frac{1}{4} & -\frac{1}{4} \\ \lambda & -\frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & -\frac{1}{4} & \lambda \\ -\frac{1}{4} & -\frac{1}{4} & \lambda \end{vmatrix} = \lambda^2 \left(\lambda - \frac{1}{2}\right) \left(\lambda + \frac{1}{2}\right)$$

得到 $\lambda_{1,2} = 0, \lambda_3 = \frac{1}{2}, \lambda_4 = -\frac{1}{2}$,故谱半径 $\rho(\boldsymbol{B}_0) = \frac{1}{2}$ 。

(2) 对于 Guass-Seidel 迭代法,

$$\boldsymbol{B}_{0} = (\boldsymbol{D} - \boldsymbol{L})^{-1} \boldsymbol{U} = \begin{pmatrix} 1 & & & \\ 0 & 1 & & \\ -\frac{1}{4} & -\frac{1}{4} & 1 & \\ -\frac{1}{4} & -\frac{1}{4} & 0 & 1 \end{pmatrix}^{-1} \begin{pmatrix} & & \frac{1}{4} & \frac{1}{4} \\ & & \frac{1}{4} & \frac{1}{4} \\ & & & \frac{1}{4} & \frac{1}{4} \\ & & & \frac{1}{8} & \frac{1}{8} \end{pmatrix}$$

求它的特征值

$$0 = |\lambda \mathbf{E} - \mathbf{B}_0| = \begin{vmatrix} \lambda & -\frac{1}{4} & -\frac{1}{4} \\ \lambda & -\frac{1}{4} & -\frac{1}{4} \\ \lambda - \frac{1}{8} & -\frac{1}{8} \end{vmatrix} = \lambda^3 \left(\lambda - \frac{1}{4}\right)$$

得到 $\lambda_{1,2,3} = 0, \lambda_4 = \frac{1}{4}$,故谱半径 $\rho(\mathbf{B}_0) = \frac{1}{4}$ 。

(3) 由于谱半径都小于 1, 所以 Jacobi 迭代法和 Guass-Seidel 迭代法均收敛。

9. 解

$$\mathbf{A} = \begin{pmatrix} 4 & -1 \\ -1 & 4 & -1 \\ & -1 & 4 \end{pmatrix} = \mathbf{D} - \mathbf{L} - \mathbf{U} = \begin{pmatrix} 4 & 4 \\ 4 & 4 \end{pmatrix} - \begin{pmatrix} 1 & 1 \\ & 1 \end{pmatrix} - \begin{pmatrix} 1 & 1 \\ & 1 \end{pmatrix}$$
$$\mathbf{b} = \begin{pmatrix} 1 \\ 4 \\ -3 \end{pmatrix}$$

对于 SOR 迭代法,松弛因子为 ω ,有

$$\boldsymbol{L}_{\omega} = (\boldsymbol{D} - \omega \boldsymbol{L})^{-1} [(1 - \omega)\boldsymbol{D} + \omega \boldsymbol{U}] = \begin{pmatrix} 4 & & \\ -\omega & 4 & \\ & -\omega & 4 \end{pmatrix}^{-1} \begin{pmatrix} 4(1 - \omega) & \omega & \\ & 4(1 - \omega) & \omega \\ & & 4(1 - \omega) \end{pmatrix}$$
$$\boldsymbol{f} = \omega (\boldsymbol{D} - \omega \boldsymbol{L})^{-1} \boldsymbol{b} = \omega \begin{pmatrix} 4 & & \\ -\omega & 4 & \\ & -\omega & 4 \end{pmatrix}^{-1} \begin{pmatrix} 1 \\ 4 \\ & -3 \end{pmatrix}$$



取迭代初值 $\mathbf{x}^{(0)} = (0,0,0)^{\mathsf{T}}$,记 $\boldsymbol{\epsilon}^{(k)} = \mathbf{x}^* - \mathbf{x}^{(k)}$,有

(1) ω = 1.03 需要 5 次迭代:

$$\mathbf{x}^{(1)} = \mathbf{L}_{\omega}\mathbf{x}^{(0)} + \mathbf{f} = (0.2575, 1.09630625, -0.49020114)^{\top} \qquad \|\boldsymbol{\epsilon}^{(1)}\|_{\infty} = 0.2425
\mathbf{x}^{(2)} = \mathbf{L}_{\omega}\mathbf{x}^{(1)} + \mathbf{f} = (0.53207386, 1.00789304, -0.49826151)^{\top} \qquad \|\boldsymbol{\epsilon}^{(2)}\|_{\infty} = 0.03207386
\mathbf{x}^{(3)} = \mathbf{L}_{\omega}\mathbf{x}^{(2)} + \mathbf{f} = (0.50107024, 1.00048646, -0.49992689)^{\top} \qquad \|\boldsymbol{\epsilon}^{(3)}\|_{\infty} = 0.00107024
\mathbf{x}^{(4)} = \mathbf{L}_{\omega}\mathbf{x}^{(3)} + \mathbf{f} = (0.50009316, 1.00002822, -0.49999493)^{\top} \qquad \|\boldsymbol{\epsilon}^{(4)}\|_{\infty} = 9.31555814 \times 10^{-5}
\mathbf{x}^{(5)} = \mathbf{L}_{\omega}\mathbf{x}^{(4)} + \mathbf{f} = (0.50000447, 1.00000161, -0.49999974)^{\top} \qquad \|\boldsymbol{\epsilon}^{(5)}\|_{\infty} = 4.47176785 \times 10^{-6}$$

(2) ω = 1 需要 6 次迭代:

$$\mathbf{x}^{(1)} = \mathbf{L}_{\omega}\mathbf{x}^{(0)} + \mathbf{f} = (0.25, 1.0625, -0.484375)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(1)}\|_{\infty} = 0.25
\mathbf{x}^{(2)} = \mathbf{L}_{\omega}\mathbf{x}^{(1)} + \mathbf{f} = (0.515625, 1.0078125, -0.49804688)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(2)}\|_{\infty} = 0.015625
\mathbf{x}^{(3)} = \mathbf{L}_{\omega}\mathbf{x}^{(2)} + \mathbf{f} = (0.50195312, 1.00097656, -0.49975586)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(3)}\|_{\infty} = 0.00195312
\mathbf{x}^{(4)} = \mathbf{L}_{\omega}\mathbf{x}^{(3)} + \mathbf{f} = (0.50024414, 1.00012207, -0.49996948)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(4)}\|_{\infty} = 0.00024414
\mathbf{x}^{(5)} = \mathbf{L}_{\omega}\mathbf{x}^{(4)} + \mathbf{f} = (0.50003052, 1.00001526, -0.49999619)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(5)}\|_{\infty} = 3.05175781 \times 10^{-5}
\mathbf{x}^{(6)} = \mathbf{L}_{\omega}\mathbf{x}^{(5)} + \mathbf{f} = (0.50000381, 1.00000191, -0.49999952)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(6)}\|_{\infty} = 3.81469727 \times 10^{-6}$$

(3) ω = 1.1 需要 6 次迭代:

$$\mathbf{x}^{(1)} = \mathbf{L}_{\omega}\mathbf{x}^{(0)} + \mathbf{f} = (0.275, 1.175625, -0.50170313)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(1)}\|_{\infty} = 0.225
\mathbf{x}^{(2)} = \mathbf{L}_{\omega}\mathbf{x}^{(1)} + \mathbf{f} = (0.57079688, 1.00143828, -0.49943416)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(2)}\|_{\infty} = 0.07079688
\mathbf{x}^{(3)} = \mathbf{L}_{\omega}\mathbf{x}^{(2)} + \mathbf{f} = (0.49331584, 0.99817363, -0.50055883)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(3)}\|_{\infty} = 0.00668416
\mathbf{x}^{(4)} = \mathbf{L}_{\omega}\mathbf{x}^{(3)} + \mathbf{f} = (0.50016617, 1.00007465, -0.49992359)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(4)}\|_{\infty} = 0.00016617
\mathbf{x}^{(5)} = \mathbf{L}_{\omega}\mathbf{x}^{(4)} + \mathbf{f} = (0.50000391, 1.00001462, -0.50000362)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(5)}\|_{\infty} = 1.46243508 \times 10^{-5}
\mathbf{x}^{(6)} = \mathbf{L}_{\omega}\mathbf{x}^{(5)} + \mathbf{f} = (0.50000363, 0.99999854, -0.50000004)^{\mathsf{T}} \qquad \|\boldsymbol{\epsilon}^{(6)}\|_{\infty} = 3.63040468 \times 10^{-6}$$

11. 证明 迭代公式

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} + \omega(\mathbf{b} - \mathbf{A}\mathbf{x}^{(k)}) = (\mathbf{E} - \omega \mathbf{A})\mathbf{x}^{(k)} + \omega \mathbf{b}$$

讨论 $L_{\omega} = E - \omega A$ 的特征值 λ , 对于任意的 $y \in \mathbb{R}^n$, 有

$$L_{\omega}y = \lambda y$$

$$(E - \omega A)y = \lambda y$$

$$y - \omega Ay = \lambda y$$
(1)

对于 A 的特征值 $0 < \alpha \le \lambda' \le \beta$,有

$$Ay = \lambda' y \tag{2}$$

结合式(1)和(2)有

$$(1 - \omega \lambda' - \lambda) \mathbf{y} = \mathbf{0}$$



考虑到对于 $\forall y \in \mathbb{R}^n$ 都成立,有

$$\lambda = 1 - \omega \lambda'$$

$$-1 = 1 - \frac{2}{\beta}\beta < \lambda < 1 - 0 = 1$$

即 $|\lambda| < 1$,则迭代法收敛。

13. 解 (1) 设
$$z^{(m)} = \begin{pmatrix} z_1^{(m)} \\ z_2^{(m)} \end{pmatrix}$$
, $b = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$, 则对于迭代方法
$$Az_1^{(m+1)} = b_1 - Bz_2^{(m)}, \quad Az_2^{(m+1)} = b_2 - Bz_1^{(m)}$$

等价为

$$\begin{pmatrix} \mathbf{A} \\ \mathbf{A} \end{pmatrix} z^{(m+1)} = \mathbf{b} - \begin{pmatrix} \mathbf{B} \\ \mathbf{B} \end{pmatrix} z^{(m)}$$

$$z^{(m+1)} = - \begin{pmatrix} \mathbf{A} \\ \mathbf{A} \end{pmatrix}^{-1} \begin{pmatrix} \mathbf{B} \\ \mathbf{B} \end{pmatrix} z^{(m)} + \begin{pmatrix} \mathbf{A} \\ \mathbf{A} \end{pmatrix}^{-1} \mathbf{b}$$

$$= - \begin{pmatrix} \mathbf{A}^{-1} \\ \mathbf{A}^{-1} \end{pmatrix} \begin{pmatrix} \mathbf{B} \\ \mathbf{B} \end{pmatrix} z^{(m)} + \begin{pmatrix} \mathbf{A}^{-1} \\ \mathbf{A}^{-1} \end{pmatrix} \mathbf{b}$$

$$= \begin{pmatrix} -\mathbf{A}^{-1} \mathbf{B} \\ -\mathbf{A}^{-1} \mathbf{B} \end{pmatrix} z^{(m)} + \begin{pmatrix} \mathbf{A}^{-1} \\ \mathbf{A}^{-1} \end{pmatrix} \mathbf{b}$$

对于迭代矩阵,它的特征值为 λ ,则对于 $\forall z = \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} \in \mathbf{R}^{2n}$,有

$$\begin{pmatrix} -A^{-1}B \\ -A^{-1}B \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \lambda \begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$$

$$\begin{cases} -A^{-1}Bz_2 = \lambda z_1 \\ -A^{-1}Bz_1 = \lambda z_2 \end{cases}$$

$$\Rightarrow \begin{cases} (A^{-1}B)^2 z_1 = \lambda^2 z_1 \\ (A^{-1}B)^2 z_2 = \lambda^2 z_2 \end{cases}$$

即
$$\rho$$
 $\begin{pmatrix} -A^{-1}B \end{pmatrix} = \rho(A^{-1}B)$, 迭代法收敛的充要条件即 $\rho(A^{-1}B) < 1$ 。
(2) 设 $z^{(m)} = \begin{pmatrix} z_1^{(m)} \\ z_2^{(m)} \end{pmatrix}$, $b = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$,则对于迭代方法
$$Az_1^{(m+1)} = b_1 - Bz_2^{(m)}, \quad Az_2^{(m+1)} = b_2 - Bz_1^{(m+1)}$$



等价为

$$\begin{pmatrix} A & \mathbf{0} \\ B & A \end{pmatrix} z^{(m+1)} = b - \begin{pmatrix} \mathbf{0} & B \\ \mathbf{0} & \mathbf{0} \end{pmatrix} z^{(m)}
z^{(m+1)} = - \begin{pmatrix} A & \mathbf{0} \\ B & A \end{pmatrix}^{-1} \begin{pmatrix} \mathbf{0} & B \\ \mathbf{0} & \mathbf{0} \end{pmatrix} z^{(m)} + \begin{pmatrix} A & \mathbf{0} \\ B & A \end{pmatrix}^{-1} b
= - \begin{pmatrix} A^{-1} & \mathbf{0} \\ -A^{-1}BA^{-1} & A^{-1} \end{pmatrix} \begin{pmatrix} \mathbf{0} & B \\ \mathbf{0} & \mathbf{0} \end{pmatrix} z^{(m)} + \begin{pmatrix} A & \mathbf{0} \\ B & A \end{pmatrix}^{-1} b
= \begin{pmatrix} \mathbf{0} & -A^{-1}B \\ \mathbf{0} & A^{-1}BA^{-1}B \end{pmatrix} z^{(m)} + \begin{pmatrix} A & \mathbf{0} \\ B & A \end{pmatrix}^{-1} b$$

对于迭代矩阵,它的特征值为 λ ,则对于 $\forall z = \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} \in \mathbf{R}^{2n}$,有

$$\begin{pmatrix} \mathbf{0} & -\mathbf{A}^{-1}\mathbf{B} \\ \mathbf{0} & \mathbf{A}^{-1}\mathbf{B}\mathbf{A}^{-1}\mathbf{B} \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \lambda \begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$$

$$\begin{cases} -\mathbf{A}^{-1}\mathbf{B}z_2 = \lambda z_1 \\ (\mathbf{A}^{-1}\mathbf{B})^2 z_2 = \lambda z_2 \end{cases}$$

$$\Rightarrow \begin{cases} (\mathbf{A}^{-1}\mathbf{B})^2 z_1 = \lambda z_1 \\ (\mathbf{A}^{-1}\mathbf{B})^2 z_2 = \lambda z_2 \end{cases}$$

即 ρ $\begin{pmatrix} \mathbf{0} & -\mathbf{A}^{-1}\mathbf{B} \\ \mathbf{0} & \mathbf{A}^{-1}\mathbf{B}\mathbf{A}^{-1}\mathbf{B} \end{pmatrix}$ = $\rho((\mathbf{A}^{-1}\mathbf{B})^2)$,迭代法的充要条件即 $\rho((\mathbf{A}^{-1}\mathbf{B})^2)$ = $\rho((\mathbf{A}^{-1}\mathbf{B}))^2 < 1$ 。 方法 (1) 的收敛速度为 $R_1 = -\ln \rho((\mathbf{A}^{-1}\mathbf{B}))$,方法 (2) 的收敛速度为 $R_2 = -2\ln \rho((\mathbf{A}^{-1}\mathbf{B}))$,即 $R_2 = 2R_1$,方法 (2) 的收敛速度是方法 (1) 的 2 倍。

14. **证明** 若矩阵
$$\mathbf{A} = \begin{pmatrix} 1 & a & a \\ a & 1 & a \\ a & a & 1 \end{pmatrix}$$
是正定的,那么

$$\begin{cases} \det \begin{pmatrix} 1 & a \\ a & 1 \end{pmatrix} = 1 - a^2 > 0 & \Rightarrow -1 < a < 1 \\ \det A = (a - 1)^2 (2a + 1) > 0 & \Rightarrow a > -\frac{1}{2} \end{cases}$$

即当 $-\frac{1}{2} < a < 1$ 时,矩阵 A 是正定的。 考察

$$\mathbf{A} = \begin{pmatrix} 1 & a & a \\ a & 1 & a \\ a & a & 1 \end{pmatrix} = \mathbf{D} - \mathbf{L} - \mathbf{U} = \begin{pmatrix} 1 & & \\ & 1 & \\ & & 1 \end{pmatrix} - \begin{pmatrix} -a & \\ -a & \\ -a & -a \end{pmatrix} - \begin{pmatrix} & -a & -a \\ & & -a \end{pmatrix}$$



在 Jacobi 迭代法中, 迭代矩阵

$$\boldsymbol{B} = \boldsymbol{D}^{-1}(\boldsymbol{L} + \boldsymbol{U}) = \begin{pmatrix} 1 & & \\ & 1 & \\ & & 1 \end{pmatrix}^{-1} \begin{pmatrix} & -a & -a \\ -a & & -a \\ -a & -a \end{pmatrix} = \begin{pmatrix} & -a & -a \\ -a & & -a \\ -a & -a \end{pmatrix}$$

考察它的特征值

$$|\lambda \mathbf{E} - \mathbf{B}| = 0 \Rightarrow \begin{vmatrix} \lambda & a & a \\ a & \lambda & a \\ a & a & \lambda \end{vmatrix} = (\lambda - a)^2 (\lambda + 2a) = 0 \Rightarrow \lambda_{1,2} = a, \lambda_3 = -\frac{1}{2a}$$

为了使迭代法收敛,有 $|\lambda| < 1$,则需要 $-\frac{1}{2} < a < \frac{1}{2}$ 。

19. **证明** (1) 由于 $(A^{T}A)^{T} = A^{T}(A^{T})^{T} = A^{T}A$,则 $A^{T}A$ 是对称矩阵。

由于 A 是非奇异矩阵,对 $\forall x \neq 0$,有 $x^{\mathsf{T}}A^{\mathsf{T}}Ax = (Ax)^{\mathsf{T}}(Ax) > 0$,所以 $A^{\mathsf{T}}A$ 是正定矩阵。 综上, $A^{\mathsf{T}}A$ 是对称正定矩阵。

(2) 由于 $A^{T}A$ 是对称矩阵,所以

$$\operatorname{cond}(\boldsymbol{A}^{\mathsf{T}}\boldsymbol{A})_{2} = \frac{|\lambda_{\max}(\boldsymbol{A}^{\mathsf{T}}\boldsymbol{A})|}{|\lambda_{\min}(\boldsymbol{A}^{\mathsf{T}}\boldsymbol{A})|}$$
(3)

另一方面,

$$\operatorname{cond}(A)_{2} = \sqrt{\frac{\lambda_{\max}(A^{\top}A)}{\lambda_{\min}(A^{\top}A)}} \tag{4}$$

综合式 (3) 和 (4) 有

$$\operatorname{cond}(A^{\mathsf{T}}A)_2 = (\operatorname{cond}(A)_2)^2$$