

$$3.1 \quad A = \begin{bmatrix} 10 & -2 & -1 \\ -2 & 10 & -1 \\ -1 & -2 & 5 \end{bmatrix} \Rightarrow \begin{cases} x_1 = 0.2x_2 + 0.1x_3 + 0.3 \\ x_2 = 0.2x_1 + 0.1x_3 + 1.5 \\ x_3 = 0.2x_1 + 0.4x_2 + 2 \end{cases}$$

$$\textcircled{1} \text{ Jacobi 法 } \Rightarrow \begin{cases} x_1^{(k+1)} = 0.2x_2^{(k)} + 0.1x_3^{(k)} + 0.3 \\ x_2^{(k+1)} = 0.2x_1^{(k)} + 0.1x_3^{(k)} + 1.5 \\ x_3^{(k+1)} = 0.2x_1^{(k)} + 0.4x_2^{(k)} + 2 \end{cases}$$

$$x^{(1)} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0.3 \\ 1.5 \\ 2 \end{pmatrix}, \quad x^{(2)} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0.2 \times 1.5 + 0.1 \times 2 + 0.3 \\ 0.2 \times 0.3 + 0.1 \times 2 + 1.5 \\ 0.2 \times 0.3 + 0.4 \times 1.5 + 2 \end{pmatrix} = \begin{pmatrix} 0.8 \\ 1.76 \\ 2.66 \end{pmatrix}$$

$$\textcircled{2} \text{ Gauss-Seidel 法 } \Rightarrow \begin{cases} x_1^{(k+1)} = 0.2x_2^{(k)} + 0.1x_3^{(k)} + 0.3 \\ x_2^{(k+1)} = 0.2x_1^{(k+1)} + 0.1x_3^{(k)} + 1.5 \\ x_3^{(k+1)} = 0.2x_1^{(k+1)} + 0.4x_2^{(k+1)} + 2 \end{cases}$$

$$x^{(1)} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0.3 \\ 0.2 \times 0.3 + 1.5 \\ 0.2 \times 0.3 + 0.4 \times 1.56 + 2 \end{pmatrix} = \begin{pmatrix} 0.3 \\ 1.56 \\ 2.684 \end{pmatrix}, \quad x^{(2)} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0.8804 \\ 1.94448 \\ 2.953872 \end{pmatrix}$$

$$3.2 \quad \Rightarrow \begin{cases} x_1 = -\frac{a_{12}}{a_{11}}x_2 + \frac{b_1}{a_{11}} \\ x_2 = -\frac{a_{21}}{a_{22}}x_1 + \frac{b_2}{a_{22}} \end{cases} \quad \text{则 } M = \begin{bmatrix} 0 & -\frac{a_{12}}{a_{11}} \\ -\frac{a_{21}}{a_{22}} & 0 \end{bmatrix}$$

$$|M - \lambda I| = 0 \Leftrightarrow \lambda^2 = \frac{a_{12}a_{21}}{a_{11}a_{22}} \quad \text{由定理知 Jacobi 法收敛的条件是}$$

$$\rho(M) < 1 \Rightarrow \text{即 } \lambda^2 < 1 \Rightarrow \left| \frac{a_{12}a_{21}}{a_{11}a_{22}} \right| < 1 \quad \text{证毕}$$

$$3.3 \quad A = \begin{bmatrix} 4 & 3 & 0 \\ 3 & 4 & -1 \\ 0 & -1 & 4 \end{bmatrix}$$

$$\textcircled{1} \text{ 最速下降法 } r^{(0)} = b - Ax^{(0)} = \begin{bmatrix} 24 \\ 30 \\ -24 \end{bmatrix} \Rightarrow p^{(0)} = r^{(0)} = \begin{bmatrix} 24 \\ 30 \\ -24 \end{bmatrix}$$

$$\alpha^{(0)} = \frac{\langle r^{(0)}, r^{(0)} \rangle}{B(r^{(0)}, r^{(0)})} = \frac{2052}{13968} = 0.1469, \quad x^{(1)} = x^{(0)} + \alpha^{(0)}p^{(0)} = \begin{bmatrix} 3.5256 \\ 4.407 \\ -3.5256 \end{bmatrix}$$

$$r^{(1)} = b - Ax^{(1)} = \begin{bmatrix} -3.325 \\ -1.732 \\ -5.489 \end{bmatrix} = p^{(1)}, \quad \alpha^{(1)} = \frac{\langle r^{(1)}, r^{(1)} \rangle}{B(r^{(1)}, r^{(1)})} = 0.2298$$

$$x^{(2)} = x^{(1)} + \alpha^{(1)}p^{(1)} = \begin{bmatrix} 2.762 \\ 4.009 \\ -4.787 \end{bmatrix}, \quad r^{(2)} = b - Ax^{(2)} = \begin{bmatrix} 0.925 \\ 0.891 \\ -0.841 \end{bmatrix} = p^{(2)}$$

$$\alpha^{(2)} = \frac{\langle r^{(2)}, r^{(2)} \rangle}{B(r^{(2)}, r^{(2)})} = 0.1485, \quad x^{(3)} = x^{(2)} + \alpha^{(2)}p^{(2)} = \begin{bmatrix} 2.899 \\ 4.141 \\ -4.912 \end{bmatrix}$$

$$\textcircled{2} \text{ 共轭梯度法 } r^{(0)} = d^{(0)} = \begin{bmatrix} 24 \\ 30 \\ -24 \end{bmatrix}, \quad \alpha^{(0)} = \frac{\langle r^{(0)}, d^{(0)} \rangle}{B(d^{(0)}, d^{(0)})} = 0.1469$$

$$x^{(1)} = x^{(0)} + \alpha^{(0)}d^{(0)} = \begin{bmatrix} 3.5256 \\ 4.407 \\ -3.5256 \end{bmatrix}, \quad r^{(1)} = b - Ax^{(1)} = \begin{bmatrix} -3.325 \\ -1.732 \\ -5.489 \end{bmatrix}$$

$$\beta^{(0)} = -\frac{B(d^{(0)}, r^{(1)})}{B(d^{(0)}, d^{(0)})} = 0.0215, \quad d^{(1)} = r^{(1)} + \beta^{(0)}d^{(0)} = \begin{bmatrix} -2.808 \\ -1.086 \\ -6.006 \end{bmatrix}$$



$$\alpha^{(3)} = \frac{\langle r^{(3)}, d^{(3)} \rangle}{B(d^{(3)}, d^{(3)})} = 0.2378, \quad x^{(3)} = x^{(2)} + \alpha^{(3)} d^{(3)} = \begin{bmatrix} -2.858 \\ 4.149 \\ -4.954 \end{bmatrix}$$

③ 与精确解  $x^* = \begin{bmatrix} 3 \\ 4 \\ -5 \end{bmatrix}$  对比  
最速下降法残差  $r = \begin{bmatrix} -0.101 \\ 0.141 \\ 0.088 \end{bmatrix}$

共轭梯度法残差  $r = \begin{bmatrix} -0.142 \\ 0.149 \\ 0.046 \end{bmatrix}$

3.4  $B = D^{-1}(D - A) = D^{-1}(L + U) = D^{-1}L + D^{-1}U$

则  $\tilde{x}^{(k)} = D^{-1}L\tilde{x}^{(k)} + D^{-1}Ux^{(k)} + D^{-1}b \quad (式1)$

$x^{(k+1)} = D^{-1}L\tilde{x}^{(k)} + D^{-1}Ux^{(k+1)} + D^{-1}b \quad (式2)$

$\Rightarrow D\tilde{x}^{(k)} = L\tilde{x}^{(k)} + Ux^{(k)} + b \Rightarrow (D - L)\tilde{x}^{(k)} = Ux^{(k)} + b$

$\Rightarrow \tilde{x}^{(k)} = (D - L)^{-1}Ux^{(k)} + (D - L)^{-1}b$

$\Rightarrow D x^{(k+1)} = L\tilde{x}^{(k)} + Ux^{(k+1)} + b \Rightarrow (D - U)x^{(k+1)} = L\tilde{x}^{(k)} + b$

$\Rightarrow x^{(k+1)} = (D - U)^{-1}L\tilde{x}^{(k)} + (D - U)^{-1}b$

$= (D - U)^{-1}L(D - L)^{-1}Ux^{(k)} + (D - U)^{-1}[L(D - L)^{-1} + I]b$

$= Mx^{(k)} + f$

则  $M = (D - U)^{-1}L(D - L)^{-1}U$

$f = (D - U)^{-1}[L(D - L)^{-1} + I]b = (D - U)^{-1}[L(D - L)^{-1} + (D - L)(D - L)^{-1}]b$

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

$= (D - U)^{-1}D(D - L)^{-1}Dg$

证毕