

$$4. a) f(x) = -12x_1^2 - 6x_1x_2 - 5x_2^2.$$

$$f(x) = \frac{1}{2} x^T Q x + b^T x + c.$$

$$i) \text{ Gradient vector } \nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} -24x_1 - 6x_2 \\ -6x_1 - 10x_2 \end{bmatrix}$$

$$ii) \text{ stationary point. } \nabla f = 0 \Rightarrow \begin{bmatrix} -24x_1 - 6x_2 \\ -6x_1 - 10x_2 \end{bmatrix} = 0.$$

$$\therefore x_1 = 0, x_2 = 0.$$

$$iii) Q = \begin{vmatrix} 24 & -6 \\ -6 & -10 \end{vmatrix} \quad b^T = 0 \quad c = 0.$$

$$q_{11} = -24 < 0, \quad |Q| = 240 - 36 > 0, \quad k \text{ even.}$$

$\therefore f(x)$ is negative definite. \therefore and Q is a negative definite matrix.

iv) Necessary condition: $\nabla f = 0$.

sufficient condition: $\nabla^2 f \succ 0 \Rightarrow \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix} \succ 0$

$$\nabla^2 f = \begin{bmatrix} -24 & -6 \\ -6 & -10 \end{bmatrix} \succ 0, \quad x^* \rightarrow \text{maximum.}$$

$$b) i) X^{k+1} = X^k - A_k^{-1} \nabla f(X^k) \quad A_k^{-1} = [\nabla^2 f(X^k)]^{-1}$$

~~$$f(x) = \frac{1}{2} x^T Q x + b^T x + c \quad \nabla f(x) = Qx + b \quad \nabla^2 f(x) = Q$$~~

$$f(x) = \frac{1}{2} x^T Q x + c^T x \quad \nabla f(x) = Qx + c \quad \nabla^2 f(x) = Q.$$

$$\nabla f(x) = 0, \quad X = -Q^{-1}c. \quad X' = X^0 - [\nabla^2 f(X^0)]^{-1} \nabla f(X^0)$$

$$= X^0 - Q^{-1}(QX^0 + c)$$

$$= X^0 - Q^{-1}QX^0 - Q^{-1}c = -Q^{-1}c.$$

$$X^2 = X' - [\nabla^2 f(X')]^{-1} \nabla f(X') = -Q^{-1}c.$$

$$X^* = -Q^{-1}c.$$

$$ii) \max f(x) = 4x_1 + 6x_2 - 2x_1^2 - 2x_1x_2 - 2x_2^2. \quad Q = \begin{bmatrix} -4 & -2 \\ -2 & -4 \end{bmatrix} \quad C^T = [4 \ 6]$$

$$f(x) = \frac{1}{2} x^T Q x + c^T x.$$

$$X^* = -Q^{-1}c. \quad Q^{-1} = \frac{\text{adj}(Q)}{\det(Q)}, \quad \det(Q) = 16 - 4 = 12, \quad \text{adj}(Q) = \begin{bmatrix} -4 & 2 \\ 2 & -4 \end{bmatrix}, \quad f(x) = \frac{7}{3}$$

$$\therefore Q^{-1} = \begin{bmatrix} -\frac{1}{3} & \frac{1}{6} \\ \frac{1}{6} & -\frac{1}{3} \end{bmatrix} \quad X^* = -\begin{bmatrix} \frac{1}{3} & \frac{1}{6} \\ \frac{1}{6} & -\frac{1}{3} \end{bmatrix} \begin{bmatrix} 4 \\ 6 \end{bmatrix} = \begin{bmatrix} -\frac{4}{3} + 1 \\ \frac{7}{3} \end{bmatrix} = \begin{bmatrix} \frac{1}{3} \\ \frac{7}{3} \end{bmatrix} \quad x_1 = \frac{1}{3}, \quad x_2 = \frac{7}{3}$$

$$\max f(x) = 4 \times \frac{7}{3} + 6 \times \frac{4}{3} - 2 \times \left(\frac{7}{3}\right)^2 - 2 \times \frac{7}{3} \times \frac{4}{3} - 2 \times \left(\frac{4}{3}\right)^2.$$

$$X^* = \begin{bmatrix} x_1^* \\ x_2^* \end{bmatrix} = \begin{bmatrix} \frac{7}{3} \\ \frac{4}{3} \end{bmatrix}$$

iii) the gradient direction refers to the direction in which the objective function increases most rapidly. the gradient of a function $f(x)$ at point X is $\nabla f(x)$. This vector points in the direction of the steepest ascent. For minimization problems, the negative gradient direction $-\nabla f(x)$ is used, as it points in the direction of the steepest descent.

$$\frac{1+5}{(1.0+5)(1.0+5)} = \frac{6}{25} \quad \frac{1+5}{(3.0+5)(1.0+5)} = \frac{6}{80} = \frac{3}{40}$$

$$= \frac{3}{50} + \frac{8}{10+5} + \frac{A}{5} = \frac{53}{50} - 25$$

$$(1.0+5)z + (1.0+5)z \cdot 8 + (1.0+5)(1.0+5)A = 1+5 \in$$

$$0.2 = \frac{1}{50.0} = A \Rightarrow 0 + 0 + 2.0 \cdot 1.0 \cdot A = 1.0 \Leftrightarrow 0 = 0 \text{ (1)}$$

$$0.8 = \frac{8}{10.0} = 0.8 \wedge 0 + (1.0)(1.0) \cdot 8 + 0 = 8.0 \Leftrightarrow 10 = 8 \text{ (2)}$$

$$0.4 = \frac{53}{50} = 1.07 \quad (1.0)(1.0) \cdot 0 + 0 + 0 = 8.0 \Leftrightarrow 5.0 = 8 \text{ (3)}$$

$$\frac{6}{50+5} \cdot 0.2 + \frac{6}{10+5} \cdot 0.8 - 0.2 = (5) \frac{7}{25} = \frac{35}{25} + \frac{48}{25} + \frac{2}{5} = \frac{85}{25} = 3.4$$

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