

数学规划 第4章作业

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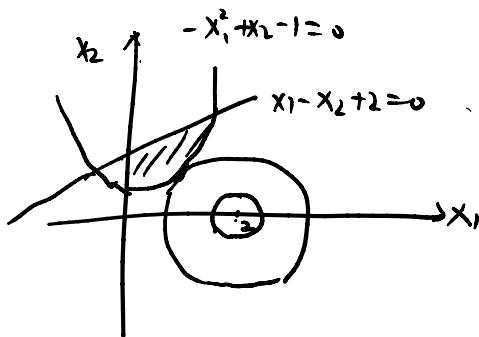
4-2 解 $\max f(x) = 3.6x_1 - 0.4x_1^2 + 1.6x_2 - 0.2x_2^2$

$$x_1 + 0.5x_2 \leq 5$$

$$x = (x_1, x_2)^T$$

4-4 (2) 解 $\min ((x_1-2)^2 + x_2^2)$

$$\text{s.t. } \begin{cases} x_2 \geq x_1^2 + 1 \\ x_2 \leq x_1 + 2 \\ x_1, x_2 \geq 0 \end{cases}$$



等高线 $(x_1-2)^2 + x_2^2 = c$.

$$\begin{cases} x_2^2 = x_1^2 + 1 \\ x_2 \leq x_1 + 2 \\ x_1, x_2 \geq 0 \end{cases}$$

$$\Rightarrow y = x_1^4 + 3x_1^2 - 4x_1 + 5 - c = 0$$

$$y' = 4x_1^3 + 6x_1 - 4$$

$$y'' = 12x_1 + b > 0$$

$$\therefore y_{\min} = y(0, 5) \approx 3.80 - c$$

$$\therefore c = 3.80$$

$$\therefore x^* = (1.554, 1.307) \quad f^* = 3.80.$$

4-5 解 (1) 梯度 $\nabla f(x) = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \frac{\partial f}{\partial x_3} \end{pmatrix} = \begin{pmatrix} 2x_1 - 4x_3 \\ 4x_2 \\ 6x_3 - 4x_1 \end{pmatrix}$

Hesse 矩阵 $\nabla(\nabla f(x)) = \begin{pmatrix} 2 & 0 & -4 \\ 0 & 4 & 0 \\ -4 & 0 & 6 \end{pmatrix}$

$x_1 \quad x_2 \quad x_3$

$$(1) \text{梯度 } \nabla f(x) = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{pmatrix} = \begin{pmatrix} 3x_2^2 + x_2 e^{x_1 x_2} \\ 6x_1 x_2 + x_1 e^{x_1 x_2} \end{pmatrix}$$

$$\text{Hesse 矩陣 } \nabla(\nabla f(x)) = \begin{pmatrix} x_2^2 e^{x_1 x_2} & 6x_2 + e^{x_1 x_2} + x_1 x_2 e^{x_1 x_2} \\ 6x_2 + e^{x_1 x_2} + x_1 x_2 e^{x_1 x_2} & 6x_1 + x_1^2 e^{x_1 x_2} \end{pmatrix}$$

$$8. \text{梯度 } \nabla f(x_1, x_2) = \begin{pmatrix} 8x_1 - 2x_2 x_1 \\ 2x_2 - x_1^2 \end{pmatrix}$$

$$\nabla(\nabla f(x_1, x_2)) = \begin{pmatrix} 8 - 2x_2 & -2x_1 \\ -2x_1 & 2 \end{pmatrix}$$

$$x^* = (0,0)^T \text{ 且 } \nabla(\nabla f(x^*)) = \begin{pmatrix} 8 & 0 \\ 0 & 2 \end{pmatrix} \text{ 不是} \quad x^* = (0,0)^T \text{ 是严格局部极小值点}.$$

$$x^* = (-2\sqrt{2}, 4)^T \text{ 且 } \nabla(\nabla f(x^*)) = \begin{pmatrix} 0 & 4\sqrt{2} \\ 4\sqrt{2} & 2 \end{pmatrix} \text{ 不是} \quad x^* = (-2\sqrt{2}, 4) \text{ 不是严格局部最小值点.}$$

$$x^* = (2\sqrt{2}, 4)^T \text{ 且 } \nabla(\nabla f(x^*)) = \begin{pmatrix} 0 & -4\sqrt{2} \\ -4\sqrt{2} & 2 \end{pmatrix} \text{ 不是} \quad x^* = (2\sqrt{2}, 4) \text{ 是}. .$$

$$4-10 \text{ (11) } \begin{cases} x \leq 4 \\ x \geq 0 \end{cases} \quad f'(x) = -3(4-x)^2, f''(x) = 6(4-x) > 0 \\ \therefore f(x) \text{ 为凸函数.}$$

$$(2) \nabla f(x) = \begin{pmatrix} 2x_1 + 2x_2 \\ 2x_1 + 6x_2 \end{pmatrix} \quad \nabla^2 f(x) = \begin{pmatrix} 2 & 2 \\ 2 & 6 \end{pmatrix} = A$$

$$|\lambda I - A| = \lambda^2 - 8\lambda + 8$$

$$\text{解得 } \lambda = 4 \pm 2\sqrt{2} > 0$$

$\therefore \nabla^2 f(x)$ 正定

$f(x)$ 为严格凸 函数.

$$(3) \nabla^2 f(x) = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = A$$

$$|\lambda I - A| = \lambda^2 - 1$$

$$\text{解得 } \lambda = \pm 1, -1 < 0$$

$\therefore f(x)$ 不是凸函数.

3. 牛顿迭代

$$\textcircled{1} \quad \lambda_1 = 3.82, \lambda_2 = 6.18 \quad b-a = 10 > \Sigma \approx 0.3$$

$$f(\lambda_1) > f(\lambda_2)$$

$$\textcircled{2} \quad a=0, b=6.18 \quad \lambda_1 = 2.36, \lambda_2 = 3.82$$

$$b-a > \Sigma \quad f(\lambda_1) < f(\lambda_2)$$

$$\textcircled{3} \quad a=0, b=3.82 \quad \lambda_1 = 1.46, \lambda_2 = -2.36$$

$$b-a > \Sigma, \quad f(\lambda_1) < f(\lambda_2)$$

$$\textcircled{4} \quad a=1.46, b=3.82 \quad \lambda_1 = 2.36, \lambda_2 = 2.92$$

$$b-a > \Sigma \quad f(\lambda_1) > f(\lambda_2)$$

$$\textcircled{5} \quad a=2.36 \quad b=3.82 \quad \lambda_1 = 2.92 \quad \lambda_2 = 3.26$$

$$b-a > \Sigma \quad f(\lambda_1) > f(\lambda_2)$$

$$\textcircled{6} \quad a=2.36 \quad b=3.26 \quad \lambda_1 = 2.70 \quad \lambda_2 = 2.92$$

$$b-a > \Sigma \quad f(\lambda_1) < f(\lambda_2)$$

$$\textcircled{7} \quad a=2.70 \quad b=3.26 \quad \lambda_1 = 2.92 \quad \lambda_2 = 3.04$$

$$b-a > \Sigma \quad f(\lambda_1) > f(\lambda_2)$$

$$\textcircled{8} \quad a=2.92 \quad b=3.26 \quad \lambda_1 = 3.04 \quad \lambda_2 = 3.14$$

$$b-a > \Sigma \quad f(\lambda_1) < f(\lambda_2)$$

$$\textcircled{9} \quad a=2.92 \quad b=3.14 \quad \lambda_1 = 3.02 \quad \lambda_2 = 3.04$$

$$b-a = 0.22 < \Sigma \quad \lambda^* = \frac{a+b}{2} = 3.03$$

$$4-14 \text{ 例 } \nabla f(x) = \begin{pmatrix} 2x_1 \\ 4x_2 \end{pmatrix} \quad A = \begin{pmatrix} 2 & 0 \\ 0 & 4 \end{pmatrix} \quad g^{(k)} = \nabla f(x^{(k)})$$

$$x^{(0)} = (4, 4)^T \quad f(x^{(0)}) = 48 \quad g^{(0)} = \nabla f(x^{(0)}) \quad \|g^{(0)}\| = 7.89 \quad \lambda_0 = \frac{g^{(0)T} \cdot g^{(0)}}{g^{(0)T} A g^{(0)}} = \frac{5}{18}.$$

$$x^{(1)} = x^{(0)} - \lambda_0 g^{(0)} = \left(\frac{16}{9}, -\frac{4}{9} \right)^T \quad f(x^{(1)}) = \frac{32}{9} \quad \|g^{(1)}\| = 2.98 \quad \lambda_1 = \frac{5}{12}$$

$$x^{(2)} = \left(\frac{8}{27}, \frac{8}{27} \right)^T \quad f(x^{(2)}) = \frac{64}{243} \quad g^{(2)} = \left(\frac{16}{27}, -\frac{9}{16} \right)^T \quad \|g^{(2)}\| = 1.23 \quad \lambda_2 = \frac{5}{8}$$

$$x^{(3)} = \left(\frac{32}{243}, \frac{-8}{243} \right)^T \quad f(x^{(3)}) = \frac{128}{6561} \quad g^{(3)} = \left(\frac{64}{243}, \frac{-32}{243} \right)^T \quad \|g^{(3)}\| = 0.29$$

$$4-15 \text{ 例 } \nabla f(x) = \begin{pmatrix} 2x_1 - 2 \\ 2x_2 \\ 18x_3 + 18 \end{pmatrix} \quad \nabla^2 f(x) = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 18 \end{pmatrix}$$

$$\nabla^2 f(x^{(0)}) = (1, 0, 0)^T$$

$$x^{(1)} = x^{(0)} - \nabla^2 f(x^{(0)})^{-1} \nabla f(x^{(0)}) = (1, 0, -1)^T$$

$$\nabla^2 f(x^{(1)}) \text{ 与 } \nabla^2 f(x^{(0)}) \text{ 相同} \Rightarrow x^{(1)} = (1, 0, -1)^T$$

$$4-16 \text{ 例 } f(x) = \frac{1}{2} x^T Q x, \quad Q = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}, \quad b = \begin{pmatrix} 2 \\ -4 \end{pmatrix}, \quad x_0 = (2, 2)^T$$

$$g(x) = \begin{pmatrix} 2x_1 - x_2 \\ -x_1 + 2x_2 - 4 \end{pmatrix}, \quad p^{(0)} = g^{(0)} = \begin{pmatrix} -4 \\ 2 \end{pmatrix}, \quad \lambda_0 = 0.36$$

$$x^{(0)} = (0.56, 0.72)^T, \quad g^{(0)} = (0.4, 0.32)^T, \quad \|g^{(0)}\| = 0.97$$

$$\beta_0 = \frac{g^{(0)T} Q p^{(0)}}{p^{(0)T} Q p^{(0)}} = 0.54, \quad p^{(0)} = -g^{(0)} + \beta_0 p^{(0)} = \begin{pmatrix} -0.616 \\ -0.772 \end{pmatrix}$$

$$\lambda_1 = \frac{g^{(1)T} g^{(1)}}{p^{(0)T} Q p^{(0)}} = 0.93$$

$$x^{(1)} = x^{(0)} + \lambda_1 p^{(0)} = \begin{pmatrix} -0.01 \\ 2.00 \end{pmatrix}$$

$$g^{(2)} = \begin{pmatrix} 0.02 \\ -0.01 \end{pmatrix}, \quad \|g^{(2)}\| = 0.015 \approx 0$$

$$\therefore x^* = x^{(2)} \in \mathbb{R}^2$$