

# 第四章

1. 用 DFP 以牛頓法 線性下代化(6) 處.

$$\min \{x_1^2 + 4x_2^2\}$$

$$\text{取 } x^{(0)} = (1, 1)^T \quad H^{(0)} = I_{2 \times 2}, \quad G = \begin{pmatrix} 2 \\ 8 \end{pmatrix}$$

$$\text{解: 簡算. } f(x) = x_1^2 + 4x_2^2 \quad g_k = (2x_1, 8x_2)^T$$

$$s_k = x_{k+1} - x_k, \quad y_k = g_{k+1} - g_k \quad \alpha_k = \frac{g_k^T H_k g_k}{d_k^T G d_k} = \frac{g_k^T H_k g_k}{g_k^T H_k G H_k g_k}$$

$$x_{k+1} = x_k - \alpha_k H_k g_k$$

$$H_{k+1} = H_k + \frac{s_k s_k^T}{s_k^T y_k} - \frac{H_k y_k y_k^T H_k}{y_k^T H_k y_k}$$

$$g_0 = (2, 8)^T \quad \alpha_0 = \frac{68}{(2, 8) \begin{pmatrix} 2 \\ 8 \end{pmatrix} \binom{2}{8}} = \frac{68}{520} = \frac{17}{130}$$

$$x_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} - \frac{17}{130} \begin{pmatrix} 1 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 8 \end{pmatrix} = \begin{pmatrix} \frac{48}{65} \\ -\frac{68}{65} \end{pmatrix}$$

$$s_0 = \begin{pmatrix} -17/65 \\ -68/65 \end{pmatrix} \quad g_1 = \begin{pmatrix} 96/65 \\ -24/65 \end{pmatrix} \quad y_0 = \begin{pmatrix} -34/65 \\ -544/65 \end{pmatrix}$$

$$H_1 = \left( \begin{array}{cc} \cancel{33537} & -\frac{526}{16705} \\ \cancel{33410} & \end{array} \right) \quad \alpha_1 = \frac{257}{520}$$

$$-\frac{526}{16705}, \quad \frac{2121}{16705}$$

$$x_2 = \begin{pmatrix} 48/65 \\ -3/65 \end{pmatrix} - \frac{257}{520} \left( \begin{array}{cc} \cancel{33537}/\cancel{33410} & -\frac{526}{16705} \\ -\frac{526}{16705} & \cancel{2121}/\cancel{16705} \end{array} \right) \begin{pmatrix} 96/65 \\ -24/65 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

而  $(0, 0)$  为  $f(x)$  最小值 (二次终止性).

故  $(0, 0)$  为原问题的最优解

## 第五章

1. 利用 KKT 条件求解下列优化问题

$$\min (x_1 - 3)^2 + (x_2 - 3)^2$$

$$\text{s.t. } 4 - x_1 - x_2 \geq 0.$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$

$$\begin{cases} L(x, \lambda) = (x_1 - 3)^2 + (x_2 - 3)^2 - \lambda \cdot (4 - x_1 - x_2) - \lambda_1 x_1 - \lambda_2 x_2. \end{cases}$$

KKT 条件为

$$\begin{cases} \frac{\partial L}{\partial x_1} = 2(x_1 - 3) + \lambda_1 - \lambda_2 = 0 \\ \frac{\partial L}{\partial x_2} = 2(x_2 - 3) + \lambda_1 - \lambda_3 = 0. \end{cases} \Rightarrow \begin{cases} x = (2, 2) \\ \lambda = (2, 0, 0). \end{cases}$$

$$4 - x_1 - x_2 \geq 0.$$

$$x_1 \geq 0.$$

$$x_2 \geq 0.$$

$$\lambda_1(4 - x_1 - x_2) = \lambda_2 x_1 = \lambda_3 x_2 = 0.$$

$$\lambda_i \geq 0, i=1, 2, 3$$

$$\nabla_{xx}^2 L(x, \lambda) = \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix} \quad \text{正定}.$$

故  $x = (2, 2)$  为最优解.

2. 考虑以下形式的约束优化问题：

$$\begin{array}{ll} \min & x_2^2 - 3x_1 \\ \text{s.t.} & x_1 + x_2 = 1 \\ & x_1 - x_2 = 0 \end{array}$$

应用外点二次罚函数法，当  $\mu \rightarrow 0$  时，求出问题的最优解与相应的拉格朗日乘子。

解： $Q(x, \mu) = x_2^2 - 3x_1 + \frac{1}{2\mu} ((x_1 + x_2 - 1)^2 + (x_1 - x_2)^2)$

$$\frac{\partial Q}{\partial x_1} = -3 + \frac{1}{2\mu} (2(x_1 + x_2 - 1) + 2(x_1 - x_2))$$

$$= -3 + \frac{1}{\mu} (x_1 - 1 + x_1) = -3 + \frac{1}{\mu} (2x_1 - 1)$$

$$\frac{\partial Q}{\partial x_2} = 2x_2 + \frac{1}{\mu} (x_1 + x_2 - 1 + x_2 - x_1)$$

$$= 2x_2 + \frac{1}{\mu} (2x_2 - 1) = (2 + \frac{2}{\mu})x_2 - \frac{1}{\mu}$$

$$\begin{cases} \frac{\partial Q}{\partial x_1} = 0 \\ \frac{\partial Q}{\partial x_2} = 0 \end{cases} \Rightarrow \begin{cases} x_1 = \frac{3\mu + 1}{2} \\ x_2 = \frac{1}{2\mu + 2} \end{cases}$$

$$\mu \downarrow 0, \text{ 则 } (x_1, x_2)^* = \left( \frac{1}{2}, \frac{1}{2} \right)$$

$$L(x, \lambda) = x_2^2 - 3x_1 - \lambda_1(x_1 + x_2 - 1) - \lambda_2(x_1 - x_2)$$

$$\begin{cases} \frac{\partial L}{\partial x_1} = -3 - \lambda_1 - \lambda_2 = 0 \\ \frac{\partial L}{\partial x_2} = 2x_2 - \lambda_1 + \lambda_2 = 0 \end{cases} \Rightarrow \begin{cases} \lambda_1 = -1 \\ \lambda_2 = -2 \end{cases}$$

$$\frac{\partial L}{\partial x_2} = 2x_2 - \lambda_1 + \lambda_2 = 0.$$

其拿 KKT 条件将  $(\lambda_1, \lambda_2)$  代入直接求解

故其最优解为  $(x_1^*, x_2^*)$ ,  $f(x^*) = -\frac{5}{4}$ , 拉格朗日乘子为  $(-1, -2)$

3. 考虑以下线性约束优化问题.

$$\begin{array}{ll} \min & 2x_1 + 3x_2 \\ \text{s.t.} & 1 - 2x_1^2 - x_2^2 \geq 0. \end{array}$$

应用对数障礙法. 当  $\mu \rightarrow 0$  时求出最优解与相应的拉格朗日乘子.

解:  $P(x, \mu) = 2x_1 + 3x_2 - \mu \log(1 - 2x_1^2 - x_2^2)$

$$\frac{\partial P}{\partial x_1} = 2 + \mu \frac{4x_1}{1 - 2x_1^2 - x_2^2}$$

$$\frac{\partial P}{\partial x_2} = 3 + \mu \frac{2x_2}{1 - 2x_1^2 - x_2^2}$$

$$\begin{cases} \frac{\partial P}{\partial x_1} = 0 \\ \frac{\partial P}{\partial x_2} = 0 \end{cases} \xrightarrow{\text{舍去一根}} \begin{cases} x_1 = -\frac{\pi}{11} \\ x_2 = -\frac{3\pi}{11} \end{cases} \quad \begin{array}{l} (\text{通过 KKT 舍去一根}), \\ \text{这里舍去因我们求的是 } P \text{ 的极} \\ \text{值).} \end{array}$$

$$L(x, \lambda) = 2x_1 + 3x_2 - \lambda(1 - 2x_1^2 - x_2^2)$$

$$\begin{cases} \frac{\partial L}{\partial x_1} = 2 + 4\lambda x_1 = 0 \\ \frac{\partial L}{\partial x_2} = 3 + 2\lambda x_2 = 0 \end{cases} \Rightarrow \lambda = \frac{\pi}{2}$$

$$\lambda > 0$$

将其代入  $x = (-\frac{\pi}{11}, -\frac{3\pi}{11})$  直接得.

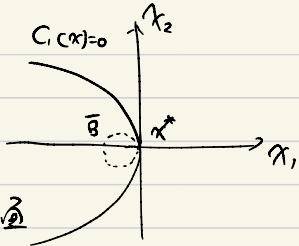
$$\text{故 } x^* = \left(-\frac{\pi}{11}, -\frac{3\pi}{11}\right) \quad f(x^*) = -\frac{\pi}{2}, \quad \lambda = \frac{\pi}{2}$$

附加题

考虑问题

$$\begin{cases} \min & f(x) = x_2 \\ \text{s.t.} & C_i(x) = -x_1 - x_2 \geq 0 \\ & x^* = (0, 0)^T \end{cases}$$

① 原点方向为  $\{d | d \in H\}$ ,  $H$  为不含  $y$  轴的左半平面



因为在  $x^*$  处的切线为  $y$  轴,  $x^*$  在  $D$  边界上,  $D$  边界光滑,  $D$  为凸集  
故作  $x^*$  处的内切圆  $B$  和存在  $x \in \bar{B}$  st.  $x + (x - x^*) \in D$ .  
而  $D$  为凸集, 故在  $x$  与  $x^*$  连线上任何一点都在  $D$  中.  
故.  $FD = \{d | d \in H\}$ ,  $H$  为不含  $y$  轴的左半平面

② 线性化原点方向.  $\nabla C(x) = (-1, 0)$   $\nabla C(x^*) = -1$

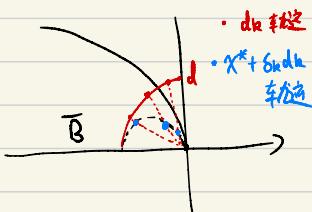
由  $d^\top \nabla C(x^*) \geq 0$  知.  $d \in H \cup Y$ .

故  $FD = \{d | d \in H \cup Y\}$ ,  $Y$  为  $y$  轴.

③ 斜割原点方向. 首先.  $FD \subseteq SF D \subseteq LFD$ .

故只需验证  $\{d | d \in Y\}$  是  $FD$  的子集.

断言其斜割原点. 在  $\bar{B}$  上取点列  $x_k \rightarrow x^*$ , 且  $x_k$  处于上半平面



此时.  $dk = (x_k - x^*) / \|x_k - x^*\|$   
 $s_k = \|x_k - x^*\|$

首先  $dk \rightarrow d$ ,  $d$  位于  $y$  轴正半轴  $d = (0, 1)^T$   
 其次.  $s_k \rightarrow 0$  因  $x_k \rightarrow x^*$ .

对  $y$  轴负半轴同理.

故.  $SFD = FD = \{d | d \in H \cup Y\}$ ,  $Y$  为  $y$  轴.