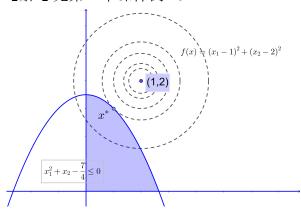
## 实用优化算法习题

0. 考虑问题

min 
$$(x_1 - 1)^2 + (x_2 - 2)^2$$
  
s.t.  $x_1^2 + x_2 - \frac{7}{4} \le 0$ ,  
 $x_1, x_2 \ge 0$ .

画出问题的可行域和等高线图, 据此求出问题的解.

【解:】见第一章课件例2.6.



1. 用斐波那契法求函数 $f(x) = x^2 - x + 2$  在区间[-1,3] 上的极小点, 要求精度为0.1, 至少迭代3 次.

【解:】先确定迭代次数:

$$\frac{3 - (-1)}{0.1} = 40.$$

写出斐波那契数列

$$F_1 = 1, F_2 = 1, F_3 = 2, F_4 = 3, F_5 = 5, F_6 = 8, F_7 = 13, F_8 = 21, F_9 = 34, F_{10} = 55.$$

故需迭代n = 10 - 1 = 9次.

第一次迭代
$$a = -1, b = 3,$$

$$x_1 = a + \frac{F_8}{F_{10}}(b-a) = -1 + \frac{21}{55} \times 4 = 0.5273, \ f_1 = f(x_1) = 1.7507;$$
  
 $x_2 = a + \frac{F_9}{F_{10}}(b-a) = -1 + \frac{34}{55} \times 4 = 1.4727, \ f_2 = f(x_2) = 2.6962.$ 

因为 $f_1 < f_2$ ,故而 $b = x_2 = 1.4727$ ,  $x_2 = x_1 = 0.5273$ ,  $f_2 = f_1 = 1.7507$ .

第二次迭代a = -1, b = 1.4727.

$$x_1 = a + \frac{F_7}{F_9}(b-a) = -1 + \frac{13}{34}(1.4727 - (-1)) = -0.0545, \ f_1 = f(x_1) = 2.0575;$$
  
 $x_2 = 0.5273, \ f_2 = 1.7507.$ 

因为 $f_1 > f_2$ ,故而 $a = x_1 = -0.0545$ ,  $x_1 = x_2 = 0.5273$ ,  $f_1 = f_2 = 1.7507$ .

第三次迭代
$$a = -0.0545$$
,  $b = 1.4727$ .

$$x_1 = 0.5273, \ f_1 = 1.7507;$$
  
 $x_2 = a + \frac{F_7}{F_8}(b - a) = 0.8909, \ f_2 = f(x_2) = 1.9028.$ 

因 $f_1 < f_2$ ,故而 $b = x_2 = 0.8909$ ,  $x_2 = x_1 = 0.5273$ ,  $f_2 = f_1 = 1.7507$ .

2. 用黄金分割法求

$$f(x) = -2x^3 + 21x^2 - 60x + 50$$

在区间[-1,4]内的最小值, 迭代三次.

【解:】第一次迭代:

$$a = -1, b = 4; x_1 = a + 0.382(b - a) = 0.91, f_1 = f(x_1) = 11.2830;$$
  
 $x_2 = 2.09, f_2 = f(x_2) = -1.9286.$ 

因为 $f_1 > f_2$ ,故而 $a = x_1 = 0.91$ , $x_1 = x_2 = 2.09$ , $f_1 = f_2 = -1.9286$ . 第二次迭代:

$$a = 0.91, b = 4; x_1 = 2.09, f_1 = -1.9286;$$
  
 $x_2 = a + 0.618(b - a) = 2.8196, f_2 = f(x_2) = 2.9448.$ 

因 $f_1 < f_2$ ,故而 $b = x_2 = 2.8196$ , $x_2 = x_1 = 2.09$ , $f_2 = f_1 = -1.9268$ . 第三次迭代:

$$a = 0.91, b = 2.8196; x_2 = 2.09, f_2 = -1.9286;$$
  
 $x_1 = a + 0.618(b - a) = 1.6395, f_1 = f(x_1) = -0.7364.$ 

因 $f_1 > f_2$ , 故而 $a = x_1 = 1.6395$ ,  $x_1 = x_2 = 2.09$ ,  $f_1 = f_2 = -1.9628$ .

3. 考虑函数

$$f(\mathbf{x}) = \frac{1}{3}x_1^3 - 16x_1 + \frac{1}{3}x_2^3 - x_2$$

写出 $\min f(x)$ 的一阶必要条件并利用该条件求f(x)的极小点.

【解:】一阶必要条件

$$\nabla f(\boldsymbol{x}) = \begin{pmatrix} x_1^2 - 16 \\ x_2^2 - 1 \end{pmatrix} = 0.$$

求解上述方程组得

$$x_1 = \pm 4, \ x_2 = \pm 1.$$

由于Hesse 矩阵

$$\nabla^2 f(\boldsymbol{x}) = \begin{pmatrix} 2x_1 & \\ & 2x_2 \end{pmatrix}$$

当 $x_1 = 4$ ,  $x_2 = 1$  时正定, 故而 $\mathbf{x} = (4,1)^T$  为问题的极小点.

## 4. 考虑下列函数

(1) 
$$f(\mathbf{x}) = 2x_1^2 + x_2^2 - 2x_1x_2 + 2x_1^3 + x_1^4$$

(2) 
$$f(\mathbf{x}) = 2x_1^3 - 3x_1^2 - 6x_1x_2(x_1 - x_2 - 1)$$
.

的所有驻点.哪些是极小点,是否是整体极小点.

【解:】(1) 函数都梯度和Hesse 矩阵分别为

$$\nabla f(\boldsymbol{x}) = \begin{pmatrix} 4x_1 - 2x_2 + 6x_1^2 + 4x_1^3 \\ 2x_2 - 2x_1 \end{pmatrix}, \ G(\boldsymbol{x}) = \begin{pmatrix} 4 + 12x_1 + 12x_1^2 & -2 \\ -2 & 2 \end{pmatrix}.$$

$$\begin{cases} x_1 = 0 \\ x_2 = 0 \end{cases} \quad \overrightarrow{\mathbb{R}} \quad \begin{cases} x_1 = -1 \\ x_2 = -1 \end{cases} \quad \overrightarrow{\mathbb{R}} \quad \begin{cases} x_1 = -\frac{1}{2} \\ x_2 = -\frac{1}{2} \end{cases} .$$

在这三个点处, Hesse 矩阵分别为

$$\begin{pmatrix} 4 & -2 \\ -2 & 2 \end{pmatrix}, \begin{pmatrix} 4 & -2 \\ -2 & 2 \end{pmatrix}, \begin{pmatrix} 1 & -2 \\ -2 & 2 \end{pmatrix}.$$

其中前两个矩阵是正定矩阵, 第三个矩阵不是正定的,也不是半正定的. 故而 $(0,0)^T$ ,  $(-1,-1)^T$  为局部极小点.

又因为f(0,0) = 0, f(-1,-1) = 0, 故而这两个点均为整体极小点.

(2) 函数的梯度和Hesse矩阵分别为

$$\nabla f(\mathbf{x}) = \begin{pmatrix} 6x_1^2 - 12x_1x_2 - 6x_1 + 6x_2^2 + 6x_2 \\ 12x_1x_2 - 6x_1^2 + 6x_1 \end{pmatrix},$$

$$G(\mathbf{x}) = \begin{pmatrix} 12x_1 - 12x_2 - 6 & 12x_2 - 12x_1 + 6 \\ 12x_2 - 12x_1 + 6 & 12x_1 \end{pmatrix}.$$

 $\diamondsuit \nabla f(\boldsymbol{x}) = 0$ , 得

$$\boldsymbol{x} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ -1 \end{pmatrix} \stackrel{\mathbf{R}}{\Rightarrow} \begin{pmatrix} -1 \\ -1 \end{pmatrix}.$$

在这四个点处的Hesse矩阵分别是

$$G = \begin{pmatrix} -6 & 6 \\ 6 & 0 \end{pmatrix}, \begin{pmatrix} 6 & -6 \\ -6 & 12 \end{pmatrix}, \begin{pmatrix} 6 & -6 \\ -6 & 0 \end{pmatrix}, \text{ All } \begin{pmatrix} -6 & 6 \\ 6 & -12 \end{pmatrix}.$$

这四个矩阵分别是鞍点矩阵, 正定矩阵, 鞍点矩阵和负定矩阵. 所以上述四个点分别为鞍点、极小点、鞍点、极大点. 其中, (1,0) 也是整体极小点.

5. 对正定二次函数 $f(\mathbf{x}) = \frac{1}{2}\mathbf{x}^T G \mathbf{x} + c^T \mathbf{x}$ , 在点 $\mathbf{x}_k$  处, 求出沿方向 $d_k$  做精确搜索的步长 $\alpha_k$ .

【解:】该二次函数的梯度为

$$\nabla f(\boldsymbol{x}) = G\boldsymbol{x} + c.$$

 $定义\varphi(\alpha) = f(\boldsymbol{x}_k + \alpha d_k), \, 则\alpha_k \, 满足$ 

$$\varphi'(\alpha) = 0,$$

也就是

$$\nabla f(\boldsymbol{x}_k + \alpha_k d_k)^T d_k = 0.$$

又

$$\nabla f(\boldsymbol{x}_k + \alpha_k d_k)^T d_k = (G(\boldsymbol{x}_k + \alpha_k d_k) + c)^T d_k$$
$$= (G\boldsymbol{x}_k + c + \alpha_k G d_k)^T d_k$$
$$= (\nabla f_k + \alpha_k G d_k)^T d_k$$
$$= \nabla f_k^T d_k + \alpha_k d_k^T G d_k = 0.$$

从而有

$$\alpha_k = -\frac{\nabla f_k^T d_k}{d_k^T G d_k}.$$

6. 用最速下降法求解

min 
$$f(\mathbf{x}) = \frac{1}{2}x_1^2 + \frac{9}{2}x_2^2$$
,

设初始点为 $(9,1)^T$ .

【见教材、课件例题】

7. 试用最速下降法求解

$$\min \ f(\mathbf{x}) = 2x_1^2 + x_2^2,$$

设初始点取为 $x_0 = (4,4)^T$ , 迭代2次, 并验证相邻两次迭代的搜索方向是正交的.

【解:】记

$$g(\mathbf{x}) = \nabla f(\mathbf{x}) = \begin{pmatrix} 2x_1 \\ 4x_2 \end{pmatrix}.$$

在初始点处 $f(\boldsymbol{x}_0)=48,\ g(\boldsymbol{x}_0)=(8,16)^T,$  搜索方向为 $d_0=-g(\boldsymbol{x}_0)=(-8,-16)$ . 考虑

$$\min_{a>0} f(\mathbf{x}_0 + \alpha d_0) = 16(3 - 20\alpha + 36\alpha_2).$$

其极小点为 $\alpha_0=5/18$ . (或者 $\alpha_0=-\frac{g_0^Td_0}{d_0^TGd_0}=\frac{5}{18}$ )

则

$$\boldsymbol{x}_1 = \boldsymbol{x}_0 + \alpha_0 d_0 = \begin{pmatrix} 16/9 \\ -4/9 \end{pmatrix}.$$

此时

$$g_1 = g(\boldsymbol{x}_1) = \begin{pmatrix} 32/9 \\ -16/9 \end{pmatrix}, d_1 = -g_1 = \begin{pmatrix} -32/9 \\ 16/9 \end{pmatrix}.$$

考虑

$$\min_{a \ge 0} f(\boldsymbol{x}_1 + \alpha d_1) = \frac{32}{81} (9 - 40\alpha + 48\alpha_2).$$

其极小点为 $\alpha_1=5/12$ . (或者 $\alpha_1=-\frac{g_1^Td_1}{d_1^TGd_1}=\frac{5}{12}$ ) 则

$$\boldsymbol{x}_2 = \boldsymbol{x}_1 + \alpha_1 d_1 = \begin{pmatrix} 8/27 \\ 8/27 \end{pmatrix}.$$

此时

$$g_2 = g(\mathbf{x}_2) = \begin{pmatrix} 16/27 \\ 32/27 \end{pmatrix}, d_2 = -g_2 = \begin{pmatrix} -16/27 \\ -32/27 \end{pmatrix}.$$

考虑

$$\min_{a \ge 0} f(\mathbf{x}_2 + \alpha d_2) = \left(\frac{8}{27}\right) (3 - 20\alpha + 36\alpha_2).$$

其极小点为 $\alpha_2=5/18$ . (或者 $\alpha_2=-\frac{g_2^Td_2}{d_2^TGd_2}=\frac{5}{18}$ ) 则

$$x_3 = x_2 + \alpha_2 d_2 = \begin{pmatrix} 32/243 \\ -8/243 \end{pmatrix}.$$

8. 用Newton 法求解

$$\min f(x) = \frac{1}{2}x_1^2 + \frac{9}{2}x_2^2.$$

【解:】Newton 法用于求解严格凸二次函数时可任选初始点,这里选 $x_0 = (1,1)^T$ . 函数都梯度和Hesse矩阵分别是

$$\nabla f(\boldsymbol{x}) = \begin{pmatrix} x_1 \\ 9x_2 \end{pmatrix}, \ G = \begin{pmatrix} 1 & 0 \\ 0 & 9 \end{pmatrix}.$$

**�** 

$$d_0 = -G^{-1}\nabla f(\boldsymbol{x}_0) = -\begin{pmatrix} 1 \\ -\frac{1}{9} \end{pmatrix} \begin{pmatrix} 1 \\ 9 \end{pmatrix} = -\begin{pmatrix} 1 \\ 1 \end{pmatrix}.$$

则

$$\boldsymbol{x}_1 = \boldsymbol{x}_0 + d_0 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

算法终止.

9. 用FR 共轭梯度法求解

min 
$$f(\mathbf{x}) = \frac{3}{2}x_1^2 + \frac{1}{2}x_2^2 - x_1x_2 - 2x_1$$
,

取初始点 $x_0 = (0,0)^T$ .

【解:】记

$$G = \begin{pmatrix} 3 & -1 \\ -1 & 1 \end{pmatrix}, \ b = \begin{pmatrix} -2 \\ 0 \end{pmatrix}.$$

则 f(x),  $\nabla f(x)$  可表示成

$$f(\boldsymbol{x}) = \frac{1}{2} \boldsymbol{x}^T G \boldsymbol{x} + b^T \boldsymbol{x}, \ \nabla f(\boldsymbol{x}) = G \boldsymbol{x} + b.$$

第一次迭代.  $\mathbf{x}_0 = (0,0)^T$ .

$$d_0 = -\nabla f(\mathbf{x}_0) = \begin{pmatrix} 2 \\ 0 \end{pmatrix}, \ \alpha_0 = -\frac{\nabla f(\mathbf{x}_0)^T d_0}{d_0^T G d_0} = \frac{1}{3}.$$

故

$$oldsymbol{x}_1 = oldsymbol{x}_0 + lpha_0 d_0 = egin{pmatrix} rac{2}{3} \\ 0 \end{pmatrix}.$$

因

$$\nabla f(\boldsymbol{x}_1) = \begin{pmatrix} 0 \\ -\frac{2}{3} \end{pmatrix} \neq 0,$$

故算法不终止.

第二次迭代.

$$\beta_0 = \frac{\|\nabla f(\boldsymbol{x}_1)\|^2}{\|\nabla f(\boldsymbol{x}_0)\|^2} = \frac{1}{9}, \ d_1 = -\nabla f(\boldsymbol{x}_1) + \beta_0 d_0 = \begin{pmatrix} \frac{2}{9} \\ \frac{2}{3} \end{pmatrix}.$$

计算步长和新迭代点

$$lpha_1 = -rac{
abla f(m{x}_1)^T d_1}{d_1^T G d_1} = rac{3}{2}, \ m{x}_2 = m{x}_1 + lpha_1 d_1 = egin{pmatrix} 1 \ 1 \end{pmatrix}.$$

因 $\nabla f(\mathbf{x}_1) = (0,0)^T$ , 故而算法终止.

函数 $f(\mathbf{x})$  的极小点为 $\mathbf{x}^* = (1,1)^T$ .

10. 考虑函数

$$f(\mathbf{x}) = 1/2x_1^2 + 1/2x_2^2$$

设初始点为 $x_0 = (1,1)^T$ ,取 $d_0 = (-1,0)^T$ ,

- (a) 沿方向 $d_0$  进行精确一维搜索得 $\alpha_0$ . 令 $\boldsymbol{x}_1 = \boldsymbol{x}_0 + \alpha_0 d_0$ , 用FR 公式求 $d_1$ .
- (b) 设G 为目标函数的Hesse 矩阵,证明 $d_0$  与 $d_1$  不是关于G 共轭的. 试说明原因。

【解:】在初始点处

$$g_0 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$
.

沿ർ0 方向进行精确搜索得最优步长为

$$\alpha_0 = -\frac{(g_0)^T d_0}{(d_0)^T G d_0} = 1.$$

则 $x_1 = x_0 + \alpha_0 d_0 = (0,1)^T$ . 计算

$$g_1 = \nabla f(\boldsymbol{x}_1) = (0, 1)^T, \ \beta_0 = \frac{\|g_1\|^2}{\|g_0\|^2} = \frac{1}{2}.$$

则

$$d_1 = -g_1 + \beta_0 d_0 = \begin{pmatrix} -\frac{1}{2} \\ -1 \end{pmatrix}.$$

由于

$$(d_0)^T G d_1 = (d_0) d_1 = \frac{1}{2} \neq 0$$

所以d<sub>0</sub>与d<sub>1</sub>不是关于

$$G = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

共轭的.

11. 用DFP 算法求解

$$\min f(x) = x_1^2 + 2x_2^2 - 2x_1x_2 - 4x_1,$$

$$\mathfrak{P} x_0 = (1,1)^T, \ H_0 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}.$$

【解:】记 $c = (-4,0)^T$ . 则目标函数可记为 $f = \frac{1}{2} \mathbf{x}^T G \mathbf{x} + c^T \mathbf{x}$ .

1)  $g_0 = \nabla f(\mathbf{x}_0) = G\mathbf{x}_0 + c = \begin{pmatrix} -4 \\ 2 \end{pmatrix}$ . 搜索方向 $d_0 = -H_0g_0 = \begin{pmatrix} 4 \\ -2 \end{pmatrix}$ . 用精确搜索,步长为

$$\alpha_0 = -\frac{g_0^T d_0}{d_0^T G d_0} = \frac{1}{4}.$$

2)  $\mathbf{x}_1 = \mathbf{x}_0 + \alpha_0 d_0 = \begin{pmatrix} 2 \\ \frac{1}{2} \end{pmatrix}, g_1 = G \mathbf{x}_1 + c = \begin{pmatrix} -1 \\ -2 \end{pmatrix}.$ 

$$s_0 = \alpha_0 d_0, y_0 = g_1 - g_0 = \begin{pmatrix} 3 \\ -4 \end{pmatrix}, H_1 = H_0 - \frac{H_0 y_0 y_0^T H_0}{y_0^T H_0 y_0} + \frac{s_0 s_0^T}{y_0^T s_0} = \begin{pmatrix} \frac{21}{25} & \frac{19}{50} \\ \frac{19}{50} & \frac{41}{100} \end{pmatrix}.$$

$$d_1 = -H_1 g_1 = \begin{pmatrix} \frac{8}{5} \\ \frac{6}{5} \end{pmatrix}$$
,用精确搜索 $\alpha_1 = \frac{5}{4}$ .

3) 
$$\mathbf{x}_2 = \mathbf{x}_1 + \alpha_1 d_1 = \begin{pmatrix} 4 \\ 2 \end{pmatrix}, g_2 = G\mathbf{x}_2 + c = \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

由于 $d_0^T G d_1 = 0$ , 故而 $d_0$ ,  $d_1$  关于G 共轭.

12. 用DFP 算法求解

$$\min f(\boldsymbol{x}) = \frac{1}{2} \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} -1 \\ 1 \end{pmatrix} + 7.$$

初始点为 $x_0 = (-1,0)^T$ ,初始矩阵为单位矩阵.

【解:】记

$$G = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix}, b = \begin{pmatrix} -1 \\ 1 \end{pmatrix}, c = 7.$$

第一次迭代.

$$\boldsymbol{x}_0 = \begin{pmatrix} -1\\0 \end{pmatrix}, \ \nabla f(\boldsymbol{x}_0) = \begin{pmatrix} -2\\1 \end{pmatrix}, \ d_0 = -\nabla f(\boldsymbol{x}_0) = \begin{pmatrix} 2\\-1 \end{pmatrix}.$$

$$\alpha_0 = -\frac{\nabla f_0^T d_0}{d_0^T G d_0} = \frac{5}{6}, \ \boldsymbol{x}_1 = \boldsymbol{x}_0 + \alpha_0 d_0 = \begin{pmatrix} \frac{2}{3}\\-\frac{5}{6} \end{pmatrix}.$$

因为

$$\nabla f(\boldsymbol{x}_1) = \begin{pmatrix} -\frac{1}{3} \\ -\frac{2}{3} \end{pmatrix} \neq 0,$$

故算法不终止.

第二次迭代.

$$s_0 = \alpha_0 d_0 = \begin{pmatrix} \frac{5}{3} \\ -\frac{5}{6} \end{pmatrix}, \ y_0 = \begin{pmatrix} \frac{5}{3} \\ -\frac{5}{3} \end{pmatrix}.$$

DFP 修正:

$$H_1 = H_0 - \frac{H_0 y_0 y_0^T H_0}{y_0^T H_0 y_0} + \frac{s_0 s_0^T}{y_0^T s_0} = \begin{pmatrix} \frac{7}{6} & \frac{1}{6} \\ \frac{2}{6} & \frac{2}{3} \end{pmatrix}.$$

计算搜索方向和步长

$$d_1 = -H_1 \nabla f_1 = \left(\frac{1}{2}, \frac{1}{2}\right), \ \alpha_1 = -\frac{\nabla f_1^T d_1}{d_1^T G d_1} = \frac{2}{3}.$$

故

$$\boldsymbol{x}_2 = \boldsymbol{x}_1 + \alpha_1 d_1 = \begin{pmatrix} 1 \\ -\frac{1}{2} \end{pmatrix}.$$

因

$$\nabla f(\boldsymbol{x}_2) = \begin{pmatrix} 0 \\ 0 \end{pmatrix},$$

故而算法终止.

13. 用DFP 算法求解

$$\min \ x_1^2 - x_1 x_2 + x_2^2 + 2x_1 - 4x_2$$

初始点取为 $x_0 = (2,2)^T$ ,初始矩阵取为单位矩阵,并验证算法所产生的两个方向关于 $G = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}$  共轭的.

【解:】记 $c=(2,-4)^T$ . 则目标函数可记为 $f=\frac{1}{2}{m x}^TG{m x}+c^T{m x}$ .

1) 
$$g_0 = \nabla f(x_0) = Gx_0 + c = \begin{pmatrix} 4 \\ -2 \end{pmatrix}$$
. 搜索方向 $d_0 = -H_0g_0 = \begin{pmatrix} -4 \\ 2 \end{pmatrix}$ . 用精确搜索,步长为

$$\alpha_0 = -\frac{g_0^T d_0}{d_0^T G d_0} = \frac{5}{14}.$$

3) 
$$\mathbf{x}_2 = \mathbf{x}_1 + \alpha_1 d_1 = \begin{pmatrix} 0 \\ 2 \end{pmatrix}, g_2 = G\mathbf{x}_2 + c = \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

由于 $d_0^T G d_1 = 0$ , 故而 $d_0$ ,  $d_1$  关于G 共轭.

14. 考虑问题

min 
$$f(x) = x_1^2 + x_2^2$$
  
s.t.  $c(x) = x_1^2 + 2x_2^2 - 1 = 0$ .

的最优性条件和极值点. (相当于用Lagrange 乘子法求极小点)

【解:】见课件第四章例题2.1.

15. 考虑优化问题

min 
$$f(x) = -(x_1 + x_2)$$
  
s.t.  $c_1(x) = 1 - x_1^2 - 4x_2^2 \ge 0$ ,  
 $x_1 \ge 0$ ,  
 $x_2 \ge 0$ .

初始点为 $x_0 = \left[\frac{1}{\sqrt{5}}, \frac{1}{\sqrt{5}}\right]$ . 试分别判断方向向量 $d_1 = [1,0]^T$ ,  $d_2 = [1,-0.5]^T$ ,  $d_3 = [0,-1]^T$  是否是初始点处的下降方向? 是否是可行方向? 是否是可行下降方向?

【解:】见课件第四章例题2.4.

16. 问题

min 
$$x_1^2 + x_2^2$$
,  
s.t.  $(x_1 - 1)^2 + x_2^2 \le 1$ ,  $x_2^2 - x_1 + 1 \le 0$  (1)

的最优解是 $x^* = (1,0)^T$ . 求点 $x^*$  处的有效集 $\mathcal{I}^*$ .

【解:】 $\mathcal{I}^* = \{2\}.$ 

17. 求问题

min 
$$f(x) = (x_1 - 1)^2 + (x_2 - 1)^2$$
,  
s.t.  $c_1(x) = 1 - x_1 - x_2 \ge 0$ ,  
 $c_2(x) = x_1 \ge 0$ ,  
 $c_3(x) = x_2 \ge 0$ .

的KKT 点.

【解:】见课件第四章例题2.6.

## 18. 求问题

min 
$$f(x) = (x_1 + x_2)^2 + 2x_1 + x_2^2$$
,  
s.t.  $x_1 + 3x_2 \le 4$ ,  
 $2x_1 + x_2 \le 3$ ,  
 $x_1, x_2 \ge 0$ 

的KKT 点.

## 【解:】问题的KKT 条件为

$$\begin{cases} 2x_1 + 2x_2 + 2 + \lambda_1 + 2\lambda_2 - \lambda_3 = 0, \\ 2x_1 + 4x_2 + 3\lambda_1 + \lambda_2 - \lambda_4 = 0, \\ 4 - x_1 - 3x_2 \ge 0, \ \lambda_1 \ge 0, \ \lambda_1(4 - x_1 - 3x_2) = 0, \\ 3 - 2x_1 - x_2 \ge 0, \ \lambda_2 \ge 0, \ \lambda_2(3 - 2x_1 - x_2) = 0, \\ x_1 \ge 0, \ \lambda_3 \ge 0, \ x_1\lambda_3 = 0, \\ x_2 \ge 0, \ \lambda_4 \ge 0, \ x_2\lambda_4 = 0. \end{cases}$$

由于

$$\lambda_3 = 2x_1 + 2x_2 + 2 + \lambda_1 + 2\lambda_2 > 2 > 0,$$

故而 $x_1=0$ .

若 $\lambda_4 > 0$ ,则 $x_2 = 0$ . 则

$$4 - x_1 - 3x_2 = 4 > 0$$
,  $3 - 2x_1 - x_2 = 3 > 0$ ,

从而 $\lambda_1 = 0$ ,  $\lambda_2 = 0$ . 那么

$$\lambda_4 = \lambda_1 + 2\lambda_2 = 0,$$

矛盾.

从而 $\lambda_4 = 0$ . 那么

$$4x_2 + 3\lambda_1 + \lambda_2 = 0.$$

这意味着

$$x_2 = 0, \ \lambda_1 = 0, \ \lambda_2 = 0.$$

所以原问题的解为

$$x_1 = 0, \ x_2 = 0,$$

相应的Lagrange 乘子为

$$\lambda_1 = 0, \ \lambda_2 = 0, \ \lambda_3 = 2, \ \lambda_4 = 0.$$

作业 优化

19. 已知约束问题

$$\min f(x) = -3x_1^2 - x_2^2 - 2x_3^2,$$
s.t.  $c_1(x) = x_1^2 + x_2^2 + x_3^2 - 3 = 0,$ 
 $c_2(x) = -x_1 + x_2 \ge 0,$ 
 $c_3(x) = x_1 \ge 0,$ 
 $c_4(x) = x_2 \ge 0,$ 
 $c_5(x) = x_3 \ge 0.$ 

试验证最优解 $x^* = (1, 1, 1)^T$  为KKT 点.

【解:】见课件第四章例题2.10.

20. 用外罚函数法求解约束优化问题

min 
$$f(x) = x_1^2 + x_2^2$$
,  
s.t.  $x_1 + x_2 - 2 = 0$ .

【解:】见课件第四章例题3.1.

21. 对问题

min 
$$x_2^2 - 3x_1$$
,  
s.t.  $x_1 + x_2 = 1$   
 $x_1 - x_2 = 0$ 

考虑外罚函数法,求出问题的局部最优解和相应的Lagrange 乘子.

【解:】构造外罚函数

$$P(\mathbf{x}; \sigma) = x_2^2 - 3x_1 + \sigma(x_1 + x_2 - 1)^2 + \sigma(x_1 - x_2)^2.$$

$$\begin{cases} \frac{\partial P}{\partial x_1} = -3 + 2\sigma(x_1 + x_2 - 1) + 2\sigma(x_1 - x_2) = 0, \\ \frac{\partial P}{\partial x_2} = 2x_2 + 2\sigma(x_1 + x_2 - 1) - 2\sigma(x_1 - x_2) = 0. \end{cases}$$

解之得

$$\begin{cases} x_1 = \frac{3+2\sigma}{4\sigma}, \\ x_2 = \frac{2\sigma}{4\sigma+2}. \end{cases}$$

$$\begin{cases} x_1 = \frac{1}{2}, \\ x_2 = \frac{1}{2}. \end{cases}$$

解之得 
$$\begin{cases} x_1 = \frac{3+2\sigma}{4\sigma}, & \lambda_1 = \frac{1}{\sigma} = \frac{1}{\sigma} \\ x_2 = \frac{2\sigma}{4\sigma+2}. & \lambda_1 = \frac{1}{\sigma} = \frac{1}{\sigma} \end{cases}$$
 
$$\begin{cases} x_1 = \frac{3+2\sigma}{4\sigma}, & \lambda_2 = \frac{1}{\sigma} \\ x_2 = \frac{2\sigma}{4\sigma+2}. & \lambda_3 = \frac{1}{\sigma} = \frac{1}{\sigma} \end{cases}$$
 
$$\begin{cases} x_1 = \frac{1}{2}, & \lambda_4 = \frac{1}{\sigma} \\ x_2 = \frac{1}{2}, & \lambda_4 = \frac{1}{\sigma} \end{cases}$$
 
$$\Rightarrow \lambda_1 = \frac{1}{\sigma}$$
 
$$\Rightarrow \lambda_1 = \frac{1}{\sigma}$$
 
$$\Rightarrow \lambda_1 = \frac{1}{\sigma}$$
 
$$\Rightarrow \lambda_2 = \frac{1}{\sigma}$$

所以, 问题的解为 $x^* = (1/2, 1/2)^T$ .

$$\nabla f - \nabla c \lambda = 0$$
 得

$$\begin{pmatrix} -\frac{1}{2} \\ 2\frac{1}{2} \end{pmatrix} - \lambda_1 \begin{pmatrix} 1 \\ 1 \end{pmatrix} - \lambda_2 \begin{pmatrix} 1 \\ -1 \end{pmatrix}_1 = 0 \implies \lambda_1 = -2$$

$$\lambda_2 = -2$$

22. 对问题

min 
$$2x_1 + 3x_2$$
  
s.t.  $1 - 2x_1^2 - x_2^2 \ge 0$ 

考虑对数障碍函数, 求出问题的解.

【解:】构造障碍函数

$$B(x;r) = 2x_1 + 3x_2 - r \ln(1 - 2x_1^2 - x_2^2).$$
 
$$\begin{cases} \frac{\partial B}{\partial x_1} = 2 + \frac{4rx_1}{1-2x_1^2 - x_2^2} = 0, \\ \frac{\partial B}{\partial x_2} = 3 + \frac{2rx_2}{1-2x_1^2 - x_2^2} = 0. \end{cases}$$
  $\Rightarrow$   $\chi_2 = 3 / 1$   $\Rightarrow$   $\chi_2 = 3 / 1$   $\Rightarrow$   $\chi_3 = 3 / 1$   $\Rightarrow$   $\chi_4 = 3 / 1$   $\Rightarrow$   $\chi_4 = 3 / 1$   $\Rightarrow$   $\chi_4 = 3 / 1$   $\Rightarrow$   $\chi_5 = 3 / 1$   $\Rightarrow$   $\chi_5 = 3 / 1$   $\Rightarrow$   $\chi_5 = 3 / 1$   $\Rightarrow$   $\chi_6 = 3 / 1$   $\Rightarrow$   $\chi_7 = 3 / 1$   $\Rightarrow$   $\chi_7$ 

解之得,

$$\begin{cases} x_1 = \frac{r + \sqrt{r^2 + 11}}{11}, \\ x_2 = \frac{3(r + \sqrt{r^2 + 11})}{11} \end{cases} \qquad \begin{cases} x_1 = \frac{r}{2}, \\ x_2 = \frac{5}{2}, \end{cases}$$

由于第一组解不在可行域内, 故而取

$$\begin{cases} x_1 = \frac{r - \sqrt{r^2 + 11}}{11}, \\ x_2 = \frac{3(r - \sqrt{r^2 + 11})}{11}. \end{cases}$$

 $r \to 0$ ,  $r = (\frac{1}{11}, \frac{3}{11})$ 

23. 写出如下不等式约束最优化问题

$$\min x_1^2 + x_2^2, s.t. x_1 + x_2 - 2 \ge 0.$$

的外罚函数.

【按定义写,略】

24. 写出如下不等式约束最优化问题

$$\min - x_1 x_2 
s.t. c_1 = -x_1 - x_2^2 + 1 \ge 0, 
c_2 = x_1 + x_2 > 0.$$

的外罚函数.

【按定义写,略】

25. 用障碍函数法

$$\min f(x_1 + x_2) = \frac{1}{3}(x_1 + 1)^3 + x_2$$
s.t.  $1 - x_1 \le 0$ ,
$$x_2 \ge 0$$
.

【解:】见课件第四章例题3.5.

26. 用内罚函数法求如下问题的最优点。

$$\min f(x) = x_1^2 + x_2^2$$

$$s.t.x_1 - x_2 + 1 \le 0$$

【解:注意不等号的方向】

构造内罚函数(障碍函数)

$$B(x,\sigma) = x_1^2 + x_2^2 - r \ln(-x_1 + x_2 - 1).$$

令

$$\begin{cases} \frac{\partial B}{\partial x_1} = -2x_1 + \frac{r}{-x_1 + x_2 - 1} = 0, \\ \frac{\partial B}{\partial x_2} = 2x_2 - \frac{r}{-x_1 + x_2 - 1} = 0. \end{cases}$$

两式相加,得 $x_1 + x_2 = 0$ . 将其回代入第一式并求解有

$$x_1 = -\frac{2 \pm \sqrt{4 + 16r^2}}{8}.$$

那么

$$x_2 = -x_1 = \frac{2 \pm \sqrt{4 + 16r^2}}{8}.$$

如果取

$$x_1 = -\frac{2 - \sqrt{4 + 16r^2}}{8},$$

则

$$x_2 = -x_1 = \frac{2 - \sqrt{4 + 16r^2}}{8}.$$

那么约束条件

$$x_1 - x_2 + 1 = \frac{\sqrt{4 + 16r^2}}{4} + 1 > 0$$

这说明该解不在可行域内部,不符合内点法的要求.从而舍弃这个解.

因此取

$$x_1 = -\frac{2 + \sqrt{4 + 16r^2}}{8},$$

球藥多?

则

$$x_2 = -x_1 = \frac{2 + \sqrt{4 + 16r^2}}{8}.$$

$$x = (-1/2, 1/2).$$

27. 用增广Lagrange 函数法求解等式约束最优化问题

$$\min x_1^2 + x_2^2,$$
s.t.  $x_1 + x_2 - 2 = 0$ .

【解:】见课件第四章例题4.2.

作业 优化

28. 用增广Lagrange 函数法求解等式约束最优化问题

$$\min f(\boldsymbol{x}) = 2x_1^2 - x_2^2 + x_1 - x_2$$
s.t.  $x_1 - x_2 = 0$ 

$$\bigwedge \left( \begin{array}{c} \begin{array}{c} \begin{array}{c} \\ \\ \end{array} \end{array} \right) \begin{array}{c} \end{array}$$

【解:】构造增广Lagrange 函数

開発している。  

$$\min f(\boldsymbol{x}) = 2x_1^2 - x_2^2 + x_1 - x_2$$
  
s.t.  $x_1 - x_2 = 0$    
 $\max f(\boldsymbol{x}) = 2x_1^2 - x_2^2 + x_1 - x_2$   
 $\max f(\boldsymbol{x}) = 2x_1^2 - x_2^2 + x_1 - x_2$   
 $\max f(\boldsymbol{x}) = 2x_1^2 - x_2^2 + x_1 - x_2$   
 $\max f(\boldsymbol{x}) = x_1 - x_2 = 0$   
 $\max f(\boldsymbol{x}) = x_1 - x_2 = 0$   
 $\max f(\boldsymbol{x}) = x_1 - x_2 = 0$   
 $\max f(\boldsymbol{x}) = x_1 - x_2 = 0$ 

$$\begin{cases} \frac{\partial M}{\partial x_1} = 4x_1 + 1 - \lambda + \sigma(x_1 - x_2) = 0, \\ \frac{\partial M}{\partial x_2} = -2x_2 - 1 + \lambda - \sigma(x_1 - x_2) = 0. \end{cases}$$

解之得

$$\begin{cases} x_1 = \frac{\lambda - 1}{4 - \sigma}, \\ x_2 = \frac{2\lambda - 2}{4 - \sigma}. \end{cases}$$

将解代入约束条件, 得 $\lambda = 1$ . 从而得问题的解为 $x^* = (0,0)^T$ .

29. 写出不等式约束最优化问题

$$\min x_1^2 + x_2^2,$$
s.t.  $x_1 + x_2 - 2 \ge 0.$ 

的增广Lagrange 函数.

【按定义写,略】

30. 用增广Lagrange 函数法求解等式约束最优化问题

$$\min f(x) = 2x_1^2 - x_2^2 + x_1 - x_2$$
  
s.t.  $x_1 - x_2 \le 0$ 

(写出增广Lagrange 函数,不需求解).

【按定义写,略】

31. 用有效集法求解

$$\min q(x) = (x_1 - 1)^2 + (x_2 - 2.5)^2$$
s.t.  $x_1 - 2x_2 + 2 \ge 0$ 

$$-x_1 - 2x_2 + 6 \ge 0$$

$$-x_1 + 2x_2 + 2 \ge 0$$

$$x_1 \ge 0$$

$$x_2 \ge 0$$

初始点为 $x_0 = (2,0)^T$ . (至少能迭代一次)

【见课件例6.2】