

Techniques in Operations Research

Exam 2016/1

①

(a)(i) we want the smallest n such that $\frac{(2-1)}{F_n} \leq 0.2 \Rightarrow \boxed{n=4}$

(ii) we want the smallest n such that $(2-1) \cdot \delta^n \leq 0.2 \Rightarrow \boxed{n=4}$

(b)

Iter	k	a_k	b_k	p_k	q_k	$f(p_k)$	$f(q_k)$
						<u>0.2</u>	<u>0</u>
1	4	1	2	1.4	1.6	0	<u>0.2</u>
2	3	1.4	2	1.6	1.8	<u>0</u>	0
3	2	1.4	1.8	1.6	1.6	<u>0</u>	

- the final interval is $[1.4, 1.6]$ (or $[1.6, 1.8]$)
- the table underlines the new computations of the obj. function.

(c) Using $\delta = 0.62$

Iter	a_k	b_k	p_k	q_k	$f(p_k)$	$f(q_k)$
					<u>0.22</u>	<u>0.02</u>
1	1	2	1.38	1.62	0.02	<u>0.16</u>
2	1.38	2	1.62	1.76	<u>0.08</u>	0.02
3	1.38	1.76	1.52	1.62		

\Rightarrow final interval $[1.52, 1.76]$

②

$$f(x,y) = 2x^3 + 2y^3 - 3x^2 + 9y^2 - 12$$

(2) First order conditions :

$$\nabla f(x,y) = \begin{bmatrix} 6x^2 - 6x \\ 6y^2 + 18y \end{bmatrix} = 0$$

$$6x(x-1) = 0 \quad \begin{cases} x=0 \\ x=1 \end{cases}$$

$$6y(y+3) = 0 \quad \begin{cases} y=0 \\ y=-3 \end{cases}$$

\Rightarrow

Stationary points

$$P_1 (0, 0)$$

$$P_2 (0, -3)$$

$$P_3 (1, 0)$$

$$P_4 (1, -3)$$

(b)

$$\nabla^2 f(x,y) = \begin{bmatrix} 12x-6 & 0 \\ 0 & 12y+18 \end{bmatrix}$$

Reason :

The Hessian has :

$$P_1 \rightarrow \nabla^2 f = \begin{bmatrix} -6 & 0 \\ 0 & 18 \end{bmatrix}$$

saddle point - (one negative and one positive eigenvalue)

$$P_2 \rightarrow \nabla^2 f = \begin{bmatrix} -6 & 0 \\ 0 & -24 \end{bmatrix}$$

maximum - (two negative eigenvalues)

$$P_3 \rightarrow \nabla^2 f = \begin{bmatrix} 6 & 0 \\ 0 & 18 \end{bmatrix}$$

minimum - (two positive eigenvalues)

$$P_4 \rightarrow \nabla^2 f = \begin{bmatrix} 6 & 0 \\ 0 & -24 \end{bmatrix}$$

saddle point - (one negative and one positive eigenvalue)

③ We want to minimise

$$(a) \quad f(\alpha, \beta) = (\alpha \beta^2 - 3)^2 + (\alpha \beta^5 - 30)^2 + (\alpha \beta^6 - 60)^2 \quad ?$$

$$(b) \quad \nabla f(\alpha, \beta) = \begin{bmatrix} 2 \cdot \beta^2 (\alpha \beta^2 - 3) + 2 \beta^5 (\alpha \beta^5 - 30) + 2 \beta^6 (\alpha \beta^6 - 60) \\ 2 \alpha \beta (\alpha \beta^2 - 3) + 10 \alpha \beta^4 (\alpha \beta^5 - 30) + 12 \alpha \beta^5 (\alpha \beta^6 - 60) \end{bmatrix}$$

$$\text{at } (\alpha, \beta) = \nabla f(2, 2) = \begin{bmatrix} 2 \cdot 4 \cdot (8 - 3) + 2 \cdot 32 (64 - 30) + 2 \cdot 64 (128 - 60) \\ 8(8 - 3) + 10 \cdot 32 (64 - 30) + 12 \cdot 64 (128 - 60) \end{bmatrix}$$

$$= \begin{bmatrix} 40 + 64 \cdot 34 + 128 \cdot 68 \\ 40 + 10 \cdot 32 \cdot 34 + 12 \cdot 64 \cdot 68 \end{bmatrix} = \begin{bmatrix} 10920 \\ 60968 \\ 63144 \end{bmatrix}$$

$$\approx \begin{bmatrix} 1 \\ 5.58 \\ 5.78 \end{bmatrix} \Rightarrow \text{steepest direction} = \begin{bmatrix} -1 \\ -5.58 \end{bmatrix}$$

$$\begin{aligned} f(t) &= (-t \cdot (-5.58t)^2 - 3)^2 + (-t \cdot (-5.58t)^5 - 30)^2 + (-t \cdot (-5.58t)^6 - 60)^2 \\ &= (-5.58^2 \cdot t^2 - 3)^2 + (5.58^6 t^6 - 30)^2 + (-5.58^6 t^7 - 60)^2 \end{aligned}$$

(9)

4 (i)

$$\begin{aligned}
 L(\underline{x}, \underline{\lambda}) = & -20x_1 - 10x_2 \\
 & + \lambda_1 (x_1^2 + x_2^2 - 1) \\
 & + \lambda_2 (-x_1) \\
 & + \lambda_3 (-x_2) \quad (1)
 \end{aligned}$$

$$(ii) \quad \left. \begin{aligned} x_1 > 0 &\Rightarrow \lambda_2 = 0 \\ x_2 > 0 &\Rightarrow \lambda_3 = 0 \end{aligned} \right\} \text{KKT} \quad (1)$$

$$\nabla L(\underline{x}, \underline{\lambda}) = \begin{bmatrix} -20 + 2\lambda_1 x_1 - \lambda_2 \\ -10 + 2\lambda_1 x_2 - \lambda_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\therefore \quad \begin{aligned} -20 + 2\lambda_1 x_1 &= 0 \\ -10 + 2\lambda_1 x_2 &= 0 \end{aligned} \quad (1) \text{ KKT}$$

$$\begin{aligned} x_1 &= \frac{10}{\lambda_1} \\ x_2 &= \frac{5}{\lambda_1} \end{aligned} \quad (1)$$

(11)

\Rightarrow Critical cone is

$$\{ (d_1, d_2) \mid 4d_1 + 2d_2 = 0 \}$$

$$\{ (d_1, -2d_1) \} \quad (2)$$

(v) For \underline{x}^* to be a local minimum

$\nabla^2 L(\underline{x}^*, \underline{\lambda}^*)$ needs to be positive definite on the critical cone. (1)

$$\nabla^2 L(\underline{x}^*, \underline{\lambda}) = \begin{bmatrix} 2\lambda_1 & 0 \\ 0 & 2\lambda_2 \end{bmatrix}$$

$$\nabla^2 L\left(\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}}, 5\sqrt{5}, 0, 0\right) = \begin{bmatrix} 10\sqrt{5} & 0 \\ 0 & 10\sqrt{5} \end{bmatrix} \quad (1)$$

which is positive definite

$\Rightarrow \underline{x}^*$ is a local min. (1) $f(\underline{x}^*) = -\frac{50}{\sqrt{5}}$ (1)

⑤

We rewrite the problem as :

$$\text{Min } f(x,y) = \frac{1}{4}x^4 - \frac{1}{2}x^2 + y^2$$

$$\text{s.t. } \begin{array}{ll} -x - a \leq 0 & \xrightarrow{a=0} -x \leq 0 \\ b - y \leq 0 & \xrightarrow{b=2} 2 - y \leq 0 \end{array}$$

$$(a) \quad P_k(x,y) = \frac{1}{4}x^4 - \frac{1}{2}x^2 + y^2 + \frac{k}{2} \left[((-x)_+)^2 + ((2-y)_+)^2 \right]$$

$$(b) \quad \nabla P_k(x,y) = \begin{bmatrix} x^3 - x - k(-x)_+ \\ 2y - k(2-y)_+ \end{bmatrix}$$

$(x)_+ = \max(0, x)$
 $= -x \quad \text{if } x < 0$

$$\nabla P_k(x,y) = 0 \quad (\text{first order necessary condition})$$

Let's assume $x < 0$:

$$x^3 - x - kx = 0 \Rightarrow x(x^2 - k - 1) = 0$$

$x^2 = k+1$

$x=0$
 or
 $x = \pm \sqrt{k+1} \quad (\text{divergent})$

therefore, the unique possibility for stationary point occurs for $x=0$, which contradicts the hypothesis

Let's assume $y \geq 2$

$$\Rightarrow 2y - k(2-y)_+ = 0 \quad \xrightarrow{=0 \text{ since } y \geq 2}$$

$$2y = 0 \Rightarrow y = 0 \quad \text{which contradicts the hypothesis}$$

therefore, stationary points only occur for

$$x \geq 0, \quad y < 2.$$

c)

$$x^3 - x = 0 \rightarrow x = 0 \quad \text{or} \quad x^2 - 1 = 0$$

$$2y - k(2-y) \rightarrow 2y - 2k + ky = 0$$

$$y(2+k) = 2k$$

$$y = \frac{2k}{2+k}$$

$$x = 1$$

(the other solution is discarded since we know $x \geq 0$)

stationary points:

$$p_1: \begin{pmatrix} 0 \\ 2 \end{pmatrix}$$

$$p_2: \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

$$x^* = \lim_{k \rightarrow \infty} x_k = 0, \quad 1$$

$$y^* = \lim_{k \rightarrow \infty} y_k = 2$$

6a) f is convex (its linear)
and all 4 constraints are
convex sets.

$$b) L(\underline{x}, \underline{\lambda}) = x_1 + 2x_2 + 3x_3 \\ + \lambda_1(x_1^2 + x_2^2 + x_3^2 - 1) \\ + \lambda_2 x_1 + \lambda_3 x_2 + \lambda_4 x_3$$

c) It is the only KKT point
as the program is convex.

$$d) L(\underline{x}^*, \underline{\lambda}^*) = -\frac{1}{\sqrt{14}} - \frac{4}{\sqrt{14}} - \frac{9}{\sqrt{14}} \\ = -\frac{14}{\sqrt{14}}.$$

$$L(\underline{x}, \underline{\lambda}^*) = x_1 + 2x_2 + 3x_3 \\ + \frac{7}{\sqrt{14}} (x_1^2 + x_2^2 + x_3^2 - 1) \\ = \frac{7}{\sqrt{14}} \left(x_1^2 + \frac{\sqrt{14}}{7} x_1 + x_2^2 + \frac{2\sqrt{14}}{7} x_2 \right. \\ \left. + x_3^2 + \frac{3\sqrt{14}}{7} x_3 - 1 \right) \\ = \frac{7}{\sqrt{14}} \left(\left(x_1 + \frac{\sqrt{14}}{14} \right)^2 - \frac{1}{14} + \left(x_2 + \frac{\sqrt{14}}{7} \right)^2 \right. \\ \left. - \frac{2}{7} + \left(x_3 + \frac{3\sqrt{14}}{14} \right)^2 - \frac{9}{14} - 1 \right)$$