

# MATH 171B HW V.

#5.1

(a) ① Substitute  $x(\alpha)$  in to  $c(x)$

then  $c(x(\alpha)) = 0$ , ~~for  $\alpha > 0$~~  for  $\alpha > 0$

②  $x(0) = (0, 0)$ , where  $(0, 0)$  is a feasible point.

③  $\frac{dx(\alpha)}{d\alpha} \Big|_{\alpha=0} = \text{~~(1, 0), (0, 1)~~$   
 $(\cos \alpha, -\sin \alpha) \Big|_{\alpha=0} = (1, 0) \neq 0$ .

All for ① ② and ③,  $x(\alpha)$  is a feasible path.

And tangent of  $x(\alpha)$  on  $(0, 0)$  is  $\frac{dx(\alpha)}{d\alpha} \Big|_{\alpha=0} = (1, 0)$ .

(b).  $f(x(\alpha)) = 3(\cos \alpha - 1) + \sin^2 \alpha + (\cos \alpha - 1)^2$

and  $f(x(0)) = 0$ .

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

$$J(x) = \begin{pmatrix} 2x_1 \\ 2(x_2 + 1) \end{pmatrix} = \nabla c(x)$$

$$\Rightarrow \nabla L(x, \lambda) = \nabla f - \nabla \lambda c(x) = \begin{pmatrix} (2-2\lambda)x_1 \\ (2-2\lambda)x_2 + 3 - 2\lambda \\ -x_1^2 - (x_2 + 1)^2 + 1 \end{pmatrix}$$

And  $\nabla_{xx}^2 L(x, \lambda) = \nabla_{xx}^2 f - \lambda \nabla_{xx}^2 c(x)$

$$= \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix} - \lambda \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$$

$$= \begin{pmatrix} 2-2\lambda & 0 \\ 0 & 2-2\lambda \end{pmatrix}$$

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(d) (i)  $(0,0)^T$  is feasible.

(ii)  $\exists \lambda = \frac{3}{2}$  s.t.  $\nabla L = 0$

(iii)  $\nabla_{xx}^2 L(x, \lambda) \Big|_{\substack{x=0 \\ \lambda=\frac{3}{2}}} = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix}$

Then  $\exists p = (1, 0)^T \in N(J(x^*))$ , s.t.

$$p^T \nabla_{xx}^2 L(x, \lambda) p \Big|_{\substack{x=0 \\ \lambda=\frac{3}{2}}} = (-1, 0) \begin{pmatrix} 1 \\ 0 \end{pmatrix} = -1 < 0.$$

Fail the necessary condition.

then  $(0,0)^T$  is not a minimizer.

# 5.2.

(a) KKT condition is 1st order necessary condition.  
i.e.  $\nabla L = 0$ .

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

$$\Rightarrow \nabla L = \begin{pmatrix} 2x_1 - \lambda \\ 4x_2 - \lambda \\ x_1 + x_2 - 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\Rightarrow \begin{cases} x_1 = \frac{2}{3} \\ x_2 = \frac{1}{3} \\ \lambda = \frac{4}{3} \end{cases}, \text{ with } x^* = \left(\frac{2}{3}, \frac{1}{3}\right)^T.$$

Then verify  $x^* = \left(\frac{2}{3}, \frac{1}{3}\right)^T$ .

Since  $x_1 = 1 - x_2$ , then objective function is

$$(1-x_2)^2 + 2x_2^2 = 3x_2^2 - 2x_2 + 1 \Rightarrow \text{minimizer is } x_2 = \frac{1}{3}$$

$\Rightarrow \left(\frac{2}{3}, \frac{1}{3}\right)^T$  is exactly an optimal point.

(b) Let  $L(x, \lambda) = x_1^3 + x_2^3 - \lambda(x_1 + x_2 - 1) = 0$

$$\nabla L = \begin{pmatrix} 3x_1^2 - \lambda \\ 3x_2^2 - \lambda \\ x_1 + x_2 - 1 \end{pmatrix} = 0 \Rightarrow \begin{cases} x_1 = \frac{1}{2} \\ x_2 = \frac{1}{2} \\ \lambda = \frac{3}{4} \end{cases} \Rightarrow x^* = \left(\frac{1}{2}, \frac{1}{2}\right)^T.$$

Then verify  $\left(\frac{1}{2}, \frac{1}{2}\right)^T$ ,

Let  $x_1 = 1 - x_2$ , then objective function

$$(1 - x_2)^3 + x_2^3 = 3x_2^2 - 3x_2 + 1$$

$$\Rightarrow \text{minimizer } x_2 = \frac{1}{2}$$

$\Rightarrow \left(\frac{1}{2}, \frac{1}{2}\right)^T$  is exactly a optimal point.