

2008 Final Exam Solution

노트 제목

2008-05-29

$$1. \quad L(A, \mu) = \frac{2}{\sqrt{3}} A_1 + A_2 + \mu \left(F \left(\frac{8}{\sqrt{3} A_1} + \frac{3}{A_2} \right) - \sigma_0 \right)$$

i) Stationary condition

$$\frac{\partial L}{\partial A_1} = \frac{2}{\sqrt{3}} - \mu F \frac{8}{\sqrt{3}} \frac{1}{A_1^2} = 0 \Rightarrow A_1 = \sqrt{4\mu F}$$

$$\frac{\partial L}{\partial A_2} = 1 - \mu F \frac{3}{A_2^2} = 0 \Rightarrow A_2 = \sqrt{3\mu F}$$

ii) feasibility condition

$$\frac{\partial L}{\partial \mu} = F \left(\frac{8}{\sqrt{3} A_1} + \frac{3}{A_2} \right) - \sigma_0 \leq 0$$

iii) orthogonality condition

$$\mu \left(F \left(\frac{8}{\sqrt{3} A_1} + \frac{3}{A_2} \right) - \sigma_0 \right) = 0$$

iv) non-negativity condition

$$\mu \geq 0$$

└ 5pt

$$\mu \neq 0$$

$$F \left(\frac{8}{\sqrt{3} A_1} + \frac{3}{A_2} \right) - \sigma_0 = 0$$

$$F \left(\frac{4}{\sqrt{3}} \frac{1}{\sqrt{\mu F}} + \frac{3}{\sqrt{3}} \frac{1}{\sqrt{\mu F}} \right) - \sigma_0 = 0 \Rightarrow \mu = \frac{49}{3} \frac{F}{\sigma_0^2}$$

$\mu = 0 \Rightarrow$ stationary condition 을 만족하지 않음.

$$\Rightarrow \mu_1 = \frac{49}{3} \frac{F}{\sigma_0^2}, \quad A_1 = \frac{14}{\sqrt{3}} \frac{F}{\sigma_0}, \quad A_2 = 7 \frac{F}{\sigma_0} \quad \text{└ 5pt}$$

- Sufficient condition

$$\underline{H}_L = \begin{bmatrix} 16\sqrt{3}\mu F A_1^{-3} & 0 \\ 0 & 6\mu F A_2^{-3} \end{bmatrix} \Rightarrow \text{P-D}$$

$F > 0, \sigma_0 > 0$
 $A_1 > 0, A_2 > 0, \mu > 0$

$$\Rightarrow \underline{d}^T \underline{H}_L \underline{d} > 0$$

└ 5pt

$$\therefore f_{\min} = \frac{49}{3} \frac{F}{\sigma_0} \quad (A_1, A_2) = \left(\frac{14}{\sqrt{3}} \frac{F}{\sigma_0}, 7 \frac{F}{\sigma_0} \right) \quad \text{└ 5pt}$$

$$\begin{aligned}
 2. \quad L &= (x_1 - 3)^2 + (x_2 + 1)^2 + \mu_1 (x_1 + x_2 - 1.5) + \mu_2 (x_1 - 1) \\
 &\triangleq L_1(x_1; \underline{\mu}) + L_2(x_2; \underline{\mu}) + L_0 \\
 &= (x_1 - 3)^2 + (\mu_1 + \mu_2)x_1 + (x_2 + 1)^2 + \mu_1 x_2 - 1.5\mu_1 - \mu_2
 \end{aligned}$$

✓ 5 pt

$$\begin{aligned}
 \phi &= \min_{x_1, x_2} L(x_1, x_2, \underline{\mu}) \\
 \begin{cases} \frac{\partial L}{\partial x_1} = 2(x_1 - 3) + (\mu_1 + \mu_2) = 0 \\ \frac{\partial L}{\partial x_2} = 2(x_2 + 1) + \mu_1 = 0 \end{cases}
 \end{aligned}$$

$$\Rightarrow \begin{cases} x_1^* = 3 - \frac{1}{2}(\mu_1 + \mu_2) \\ x_2^* = -1 - \frac{1}{2}\mu_1 \end{cases} \quad \text{✓ 5 pt}$$

$$\phi(\underline{\mu}) = -\frac{1}{2}\mu_1^2 - \frac{1}{4}\mu_2^2 - \frac{1}{2}\mu_1\mu_2 + \frac{1}{2}\mu_1 + 2\mu_2 \quad \text{✓ 10 pt}$$

i) $\mu_1 > 0, \mu_2 > 0$

$$\begin{aligned}
 \frac{\partial \phi}{\partial \mu_1} = -\mu_1 - \frac{1}{2}\mu_2 + \frac{1}{2} = 0 \\
 \frac{\partial \phi}{\partial \mu_2} = -\frac{1}{2}\mu_2 - \frac{1}{2}\mu_1 + 2 = 0
 \end{aligned}
 \Rightarrow \mu_1 = -3, \mu_2 = 7 \quad \text{✓ 6 pt}$$

ii) $\mu_2 > 0, \mu_1 = 0$ (i.e. $g_1 \leq 0$)

$$\mu_2 = 4, \quad x_1^* = 1, \quad x_2^* = -1 \quad g_1 = -1.5 \leq 0 \quad \text{✓ 6 pt}$$

iii) $\mu_1 > 0, \mu_2 = 0$ (i.e. $g_2 \leq 0$)

$$\mu_1 = \frac{1}{2}, \quad x_1^* = \frac{11}{4}, \quad x_2^* = -\frac{5}{4}, \quad g_2 = \frac{7}{4} > 0$$

not a solution ✓ 6 pt

Sufficient Condition

$$\frac{\partial^2 \phi}{\partial \mu_1 \partial \mu_2} \bigg|_{\underline{\mu}^*} = \frac{1}{4} \begin{bmatrix} -4 & -2 \\ -2 & -2 \end{bmatrix} \Rightarrow \text{N-D} \quad \text{✓ 4 pt}$$

$$\underline{\mu}^* = (0, 4) \quad \phi_{\max} = 4$$

$$\underline{x}^* = (1, -1) \quad f_{\min} = 4 \quad \text{✓ 8 pt}$$

$$3. L(x_1, x_2, \mu) = x_1^2 + x_2^2 + \mu(-(x_1-1)^3 + x_2^2)$$

(a) KKT condition

i) stationary condition

$$\frac{\partial L}{\partial x_1} = 2x_1 - \mu 3(x_1-1)^2 = 0$$

$$\frac{\partial L}{\partial x_2} = 2x_2 + \mu 2x_2 = 0$$

ii) feasibility condition

$$\frac{\partial L}{\partial \mu} = -(x_1-1)^3 + x_2^2 \leq 0$$

iii) orthogonality condition

$$\mu(-(x_1-1)^3 + x_2^2) = 0$$

iv) non-negativity

$$\mu \geq 0$$

✓ 10 pt

(b) $\mu \neq 0$

$$2x_2(\mu+1) = 0 \quad \mu > 0 \quad \text{이므로}$$

$$x_2 = 0 \quad x_1 = 1 \quad \text{feasibility}$$

$$\rightarrow 2x_1 - \mu 3(x_1-1)^2 = 2 \neq 0 \quad \text{모순}$$

$$\mu = 0$$

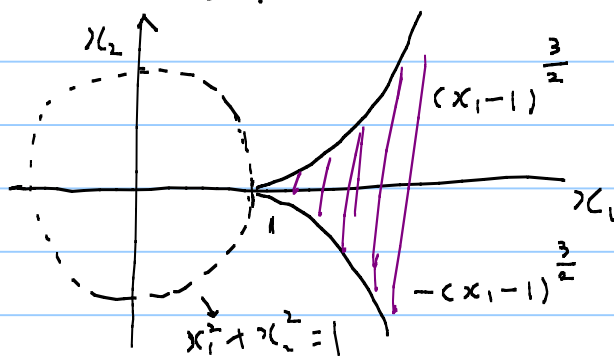
$$x_1 = 0, \quad x_2 = 0$$

$$\rightarrow -(-1)^3 + 0 \geq 0 \quad \text{모순}$$

✓ 10 pt

KKT condition 으로부터 solution 을 구할 수 있다.

$$x_2^2 \leq (x_1-1)^3 \Rightarrow -(x_1-1)^{\frac{3}{2}} \leq x_2 \leq (x_1-1)^{\frac{3}{2}}$$



✓ 5 pt

$(x_1, x_2) = (1, 0)$ 에서 $f_{\min} = 1$

KKT condition 으로 풀 수 없는 이유

\rightarrow CQ does not hold at $(1, 0)$ ✓ 5 pt

$$4. \quad SS_T = \sum_{i=1}^n (y_i - \bar{y})^2 = \sum_{i=1}^n \left\{ (y_i - \hat{y}_i) + (\hat{y}_i - \bar{y}) \right\}^2$$

$$= \underbrace{\sum_{i=1}^n (y_i - \hat{y}_i)^2}_{SS_E} + \underbrace{\sum_{i=1}^n (\hat{y}_i - \bar{y})^2}_{SS_R} + 2 \sum_{i=1}^n (y_i - \hat{y}_i)(\hat{y}_i - \bar{y})$$
✓ 5 pt

$$e_i = y_i - \hat{y}_i$$

$$\Rightarrow \sum_{i=1}^n (\hat{y}_i - \bar{y})(y_i - \hat{y}_i) = \sum_{i=1}^n (\hat{y}_i - \bar{y})e_i$$
✓ 5 pt

$$L = \underline{e}^T \underline{e} = (\underline{y} - \underline{X}\underline{\beta})^T (\underline{y} - \underline{X}\underline{\beta})$$

$$\begin{aligned} (\underline{\beta}^T \underline{X}^T \underline{y} = \underline{y}^T \underline{X} \underline{\beta}) &= \underline{y}^T \underline{y} - \underline{\beta}^T \underline{X}^T \underline{y} - \underline{y}^T \underline{X} \underline{\beta} + \underline{\beta}^T \underline{X}^T \underline{X} \underline{\beta} \\ \text{scalar} &= \underline{y}^T \underline{y} - 2 \underline{\beta}^T \underline{X}^T \underline{y} + \underline{\beta}^T \underline{X}^T \underline{X} \underline{\beta} \end{aligned}$$

to minimize error L

$$\frac{\partial L}{\partial \underline{\beta}} \Big|_{\underline{\beta} = \underline{b}} = -2 \underline{X}^T \underline{y} + 2 \underline{X}^T \underline{X} \underline{b} = 0$$

$$\rightarrow (\underline{X}^T \underline{X}) \underline{b} = \underline{X}^T \underline{y} \quad \text{----- (1)}$$

$$\begin{bmatrix} n & \sum_{i=1}^n x_{i1} & \dots & \sum_{i=1}^n x_{ik} \\ \sum_{i=1}^n x_{i1} & \sum_{i=1}^n x_{i1}^2 & \dots & \sum_{i=1}^n x_{i1}x_{ik} \\ \vdots & \vdots & \ddots & \vdots \\ \sum_{i=1}^n x_{ik} & \sum_{i=1}^n x_{ik}x_{i1} & \dots & \sum_{i=1}^n x_{ik}^2 \end{bmatrix} \begin{Bmatrix} b_0 \\ b_1 \\ \vdots \\ b_k \end{Bmatrix} = \begin{Bmatrix} \sum_{i=1}^n y_i \\ \sum_{i=1}^n x_{i1} y_i \\ \vdots \\ \sum_{i=1}^n x_{ik} y_i \end{Bmatrix}$$

$$\Rightarrow nb_0 + \sum_{j=1}^k b_j \sum_{i=1}^n x_{ij} = \sum_{i=1}^n y_i \quad \text{----- (2)}$$

$$\hat{\underline{y}} = \underline{X} \underline{b} \quad \text{----- (3)}$$

✓ 20 pt

$$\sum_{i=1}^n (\hat{y}_i - \bar{y}) e_i = 0 \quad \text{증명}$$

$$\begin{aligned} \sum_{i=1}^n \hat{y}_i e_i &= \hat{\beta}^T e = (X \hat{\beta})^T (y - \hat{y}) \quad \leftarrow (3) \\ &= \hat{\beta}^T X^T y - \hat{\beta}^T X^T X \hat{\beta} \\ &= \hat{\beta}^T (X^T y - X^T X \hat{\beta}) \quad \leftarrow (1) \\ &= 0 \end{aligned}$$

$$\begin{aligned} \sum_{i=1}^n e_i &= \sum_{i=1}^n (y_i - \hat{y}_i) \quad \checkmark \text{10 pt} \\ &= \sum_{i=1}^n \left\{ y_i - \left(b_0 + \sum_{j=1}^k b_j x_{ij} \right) \right\} \\ &= \sum_{i=1}^n y_i - \left(n b_0 + \sum_{j=1}^k b_j \sum_{i=1}^n x_{ij} \right) \quad \leftarrow (2) \end{aligned}$$

$$\begin{aligned} &= 0 \\ \Rightarrow \sum_{i=1}^n \bar{y} e_i &= 0 \Rightarrow \sum_{i=1}^n (\hat{y}_i - \bar{y}) e_i = 0 \quad \checkmark \text{10 pt} \end{aligned}$$

$$\Rightarrow SS_T = SS_R + SS_E$$