

Subject:

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$$L(x, \lambda) = f_0(x) + \sum_{i=1}^m \lambda_i f_i(x) \quad f_i(x): \text{Linear} \quad (1.0)$$

$$= 4x_1 + 3x_2 + \lambda_1^T (-5x_1 - x_2 + 11) + \lambda_2^T (-2x_1 + x_2 + 8) + \lambda_3^T (-x_1 - 2x_2 + 7) + \lambda_4^T (-x_1) + \lambda_5^T (-x_2)$$

$$g(\lambda) = \inf_x (L(x, \lambda)) = \inf_x (L(x_1, x_2, \lambda_1, \lambda_2, \lambda_3, \lambda_4, \lambda_5))$$

$$\frac{\partial L}{\partial x_1} = 4 - 5\lambda_1 - 2\lambda_2 - \lambda_3 - \lambda_4 = 0 \quad \times 3 \rightarrow 12 - 15\lambda_1 - 6\lambda_2 - 3\lambda_3 - 3\lambda_4 = 0$$

$$\frac{\partial L}{\partial x_2} = 3 - \lambda_1 - \lambda_2 - 2\lambda_3 - \lambda_5 = 0 \quad \times 4 \rightarrow 12 + 4\lambda_1 + 4\lambda_2 + 8\lambda_3 + 4\lambda_5 = 0$$

$$\oplus \rightarrow -11\lambda_1$$

A

$$g = \inf \left\{ (4 - 5\lambda_1 - 2\lambda_2 - \lambda_3 - \lambda_4)^T x_1 + (3 - \lambda_1 - \lambda_2 - 2\lambda_3 - \lambda_5)^T x_2 + 11\lambda_1 + 8\lambda_2 + 7\lambda_3 \right\}$$

B

$$= B + \inf A$$

$$\Rightarrow \text{maximize } B$$

$$\text{s.t. } \begin{cases} 4 - 5\lambda_1 - 2\lambda_2 - \lambda_3 - \lambda_4 = 0 \\ 3 - \lambda_1 - \lambda_2 - 2\lambda_3 - \lambda_5 = 0 \end{cases}$$

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$$P_j \geq 0$$

or

$$\min \sum_{j=1}^n P_j$$

$$\text{s.t. } \sum_{j=1}^n g_{ij} \cdot P_j \geq P, \quad i=1, \dots, m$$

$$P_j \geq 0, \quad j=1, \dots, n$$

KKT

1) Primal constraints: $f_i(u) \leq 0, \quad h_i(u) = 0$

$$\hookrightarrow \begin{cases} g_{ij} \cdot P_j \geq P, & i=1, \dots, m, j=1, \dots, n \\ P_j \geq 0, & j=1, \dots, n \end{cases}$$

2) Dual constraints: $\lambda_{ij} \geq 0 \quad \checkmark$

3) Complementary slackness: $\lambda_i f_i(u) = 0$

$$\hookrightarrow \lambda_{ij} \left(\sum_{j=1}^n g_{ij} \cdot P_j - P \right) = 0, \quad i=1, \dots, m, j=1, \dots, n \quad \left(\sum_{j=1}^n g_{ij} \cdot P_j \geq P \right)$$

$$4) \nabla L = 0, \quad L = \sum_{j=1}^n P_j + \sum_{i=1}^m \sum_{j=1}^n \lambda_{ij} \cdot \left(P - \sum_{j=1}^n g_{ij} \cdot P_j \right)$$

$$\hookrightarrow \frac{\partial L}{\partial P_j} = 1 - \sum_{i=1}^m \lambda_{ij} \cdot g_{ij} = 0 \quad \checkmark$$

TISS

401133004 عین سہ ماہی کے لیے - اس کے لیے

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$$P(x_i = k) = \frac{e^{-\mu_i} \mu_i^k}{k!} \quad \text{و } y = (y_1, y_2, \dots, y_m) \quad (3 \text{ نکات})$$

$$y_j = \sum_{i=1}^n y_{ji}$$

$$l(a, b) = l(\mu | y) = \sum_{j=1}^m \prod_{i=1}^n \log \left(\frac{e^{-\mu_i} \mu_i^{y_{ji}}}{y_{ji}!} \right) \quad (15 \text{ نکات})$$

$$\Rightarrow l(\mu | y) = \sum_{i=1}^m \sum_{j=1}^n (-\mu_i + y_{ji} \log \mu_i - \log(y_{ji}!))$$

$$\hookrightarrow \max l(\mu | y)$$

$$\text{s.t.} \begin{cases} \sum_{j=1}^m p_{ji} = 1 \\ p_{ji} \geq 0 \end{cases}$$

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سوال (4) تابع هزینه لایسم (برای 15)

$$p = \text{Prob}(y=1) = \frac{\exp(a^T u + b + v)}{1 + \exp(a^T u + b + v)}, \quad 1-p = \text{Prob}(y=0)$$

$$l(a, b) = \log \prod_{i=1}^q P_i \prod_{i=q+1}^m (1 - P_i)$$

$$= \sum_{i=1}^q \log P_i + \sum_{i=q+1}^m \log(1 - P_i)$$

$$= \sum_{i=1}^q \log \frac{\exp(a^T u_i + b + v)}{1 + \exp(a^T u_i + b + v)} + \sum_{i=q+1}^m \log \frac{1}{1 + \exp(a^T u_i + b + v)}$$

$$l(a, b) = \sum_{i=1}^q (a^T u_i + b + v) - \sum_{i=1}^m \log(1 + \exp(a^T u_i + b + v))$$

قسم اول ≥ 1
قسم دوم

max $l(a, b)$

min $-l(a, b) \rightarrow \text{Convex} \checkmark$

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$$\min \sum_{i=1}^m \phi(r_i) \quad x \in \mathbb{R}^n, A \in \mathbb{R}^{m \times n}$$

$$\text{s.t. } r = Ax - b$$

$$L(x, r, \lambda) = \sum_{i=1}^m \phi(r_i) + \lambda^T (Ax - b - r)$$

$$\hookrightarrow g(\lambda) = \inf_x L(x, r, \lambda) \quad , \quad \text{dual: } \phi^*(y) = \sup_x \{x^T y - \phi(x)\}$$

$$\Rightarrow g(\lambda) = - \sum_{i=1}^m \phi^*(-\lambda_i) - b^T \lambda$$

$$\hookrightarrow \begin{array}{ll} \max & g(\lambda) \\ \text{s.t.} & \lambda \geq 0 \end{array} \quad \checkmark$$

$$b) \quad \phi(x) = \begin{cases} 0, & |x| \leq \frac{1}{2} \\ |x| - \frac{1}{2}, & |x| > \frac{1}{2} \end{cases} \quad \rightarrow \quad \phi^*(y) = \begin{cases} 0, & |y| \leq \frac{1}{2} \\ |y| - \frac{1}{2}, & |y| > \frac{1}{2} \end{cases}$$

$$\Rightarrow g(\lambda) = - \sum_{i=1}^m \phi^*(-\lambda_i) - b^T \lambda$$

$$= -b^T \lambda - \sum_{i=1}^m (-\lambda_i + \frac{1}{2})$$

$$\hookrightarrow \max \quad -b^T \lambda + \sum \lambda_i - \frac{1}{2}$$

$$\text{s.t. } \lambda \geq 0$$

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$$H = \{x \in \mathbb{R}^n \mid a^T x + b \geq 0\}$$

(9 د)

minimize $f(x) = \frac{1}{2} \|x\|^2$

s.t : $g(x) = a^T x + b \geq 0$

QP

جواب KKT

$$L(x, \lambda) = \frac{1}{2} \|x\|^2 - \lambda(a^T x + b)$$

KKT: $\lambda \geq 0, A^T \lambda = 0$

$A x^* = b$

min $f_0(x)$

s.t. $f_i(x) \leq 0, i=1, \dots, m$

(v.d)

KKT $\lambda_i f_i(x^*) = 0$

$$\Rightarrow \lambda_0 \nabla f_0(x^*)^T (x - x^*) + \sum_{i=1}^m \lambda_i \nabla f_i(x^*)^T (x - x^*) = 0 \quad (I)$$

$$\Rightarrow \nabla f_0(x^*)^T + \sum_{i=1}^m \lambda_i \nabla f_i(x^*)^T = 0$$

$$\Rightarrow \nabla f_0(x^*)^T (x - x^*) + \sum_{i=1}^m \lambda_i \nabla f_i(x^*)^T (x - x^*) = 0 \quad (II)$$

(I, II)

$$\Rightarrow \lambda_0 \nabla f_0(x^*)^T (x - x^*) + \sum_{i=1}^m \lambda_i \nabla f_i(x^*)^T (x - x^*) = \nabla f_0(x^*)^T (x - x^*) + \sum_{i=1}^m \lambda_i \nabla f_i(x^*)^T (x - x^*)$$

$$\Rightarrow (x - x^*)^T \left(\nabla f_0(x^*)^T + \sum_{i=1}^m \lambda_i \nabla f_i(x^*)^T \right) + \sum_{i=1}^m (\lambda_i \nabla f_i(x^*)^T) (x - x^*) = 0$$

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$$\Rightarrow \nabla f(x^*)^T (x - x^*) + \sum_{i=1}^m (\lambda_i \nabla f_i(x^*)^T) (x - x^*) \geq \dots \quad (\text{Feasible})$$

⊙

$$\text{KKT: } f_i(x^*) \geq 0 \quad \lambda_i \geq 0$$

$$\hookrightarrow \nabla f(x^*)^T (x - x^*) + \sum_{i=1}^m (\lambda_i \nabla f_i(x^*)^T) (x - x^*) = \nabla f(x^*)^T (x - x^*) \geq 0 \quad \checkmark$$

feasible set is S_1

~~⊙~~

$$\phi(x) = - \sum_{i=1}^m \log(-f_i(x))$$

(1) \checkmark

$$\tilde{\phi}(x) = \sum_{i=1}^m t \log(-f_i(x)) - t \log(x^T x - R^2)$$

$$\rightarrow \nabla^2 (t f_0(x) + \tilde{\phi}(x)) = t \nabla^2 f_0(x) + \nabla^2 \tilde{\phi}(x) \succcurlyeq \alpha I$$

$$\nabla^2 \phi_i(x) = \frac{1}{(f_i(x))^2} \nabla f_i(x) \nabla f_i(x)^T - \frac{1}{f_i(x)} \nabla^2 f_i(x)$$

↓ \checkmark
Sink