

نقطة استقرار : $P(\frac{1}{4}) = \begin{pmatrix} \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{2} \end{pmatrix}$

$$\begin{pmatrix} 1 & 0 \\ 1 & 5 \\ 1 & 10 \\ 1 & 15 \end{pmatrix} \begin{pmatrix} \ln a \\ b \end{pmatrix} = \begin{pmatrix} \ln(227) \\ \ln(237) \\ \ln(249) \\ \ln(262) \end{pmatrix}$$

$$X^* = (A^T A)^{-1} A^T b \rightarrow \bar{A} A = \begin{pmatrix} 4 & 30 \\ 20 & 350 \end{pmatrix}$$

$$\bar{A} b = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & 5 & 10 & 15 \end{pmatrix} \begin{pmatrix} 5.42 \\ 5.46 \\ 5.51 \\ 5.56 \end{pmatrix} = \begin{pmatrix} 21.97 \\ 166.03 \end{pmatrix}$$

$$\begin{pmatrix} 4 & 30 \\ 20 & 350 \end{pmatrix} \begin{pmatrix} \ln a \\ b \end{pmatrix} = \begin{pmatrix} 21.97 \\ 166.03 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 5.42 \\ 0.0095 \end{pmatrix}$$

$$\ln a = 5.42 \rightarrow a = 226.5$$

$$P(t) = 226.5 e^{0.0092t}$$

$$P(30) = 226.5 e^{0.0095 \times 30} = 261.19$$

$$f(x, \theta) = (1 - \theta)^{x-1} \theta$$

سوال ۳

$$\text{Likelihood: } L(\theta) = (1 - \theta)^{x_1-1} \theta (1 - \theta)^{x_2-1} \theta \dots$$

$$(1 - \theta)^{x_n-1} \theta = \theta^n (1 - \theta)^{\sum_{i=1}^n x_i - n}$$

$$\text{log likelihood: } \ln L(\theta) = n \ln \theta + \sum_{i=1}^n x_i - n \cdot \ln(1 - \theta)$$

$$\frac{d}{d\theta} \ln L(\theta) = \frac{n}{\theta} - \frac{\sum_{i=1}^n x_i - n}{1 - \theta} = 0 \rightarrow \theta = \frac{n}{\sum_{i=1}^n x_i} = \frac{1}{\bar{x}}$$

$$\begin{cases} P(0) : P(y=0) = 1 - Q(a^T u + b) = P(-a^T u - b) \\ P(1) : P(y=1) = Q(a^T u + b) \end{cases}$$

سوال ۴

$$Q(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{\frac{t^2}{2}} dt$$

$$\text{log likelihood: } \sum_{y_i=1} \log Q(a^T u_i + b) + \sum_{y_i=0} \log Q(-a^T u_i - b)$$

در ادامه به دست می آید.

$$\begin{aligned} \min & \sum_{i=1}^n x_i \log(x_i) \\ \text{s.t.} & Ax \leq b \\ & [1]^T x = 1 \end{aligned}$$

سوال ۵

Linear program: $\sum x_i \log(x_i) + \lambda^T (Ax - b) + r (1^T x - 1)$

$$\begin{aligned} \mathcal{J}(\lambda, r) &= \inf_x L_p = -\lambda^T b - r + \inf (\sum x_i \log x_i \\ &+ (\lambda^T A + r \mathbf{1}^T) x_i) \end{aligned}$$

$$= -\sup (-A^T \lambda - r \mathbf{1})^T x - \sum_{i=1}^n x_i \log x_i$$

$$\rightarrow y = -A^T \lambda - r \mathbf{1}$$

$$f^{\infty}(-A^T \lambda - r \mathbf{1}) = f^{\infty}(y) = -\sup (y^T A - f_{\infty})$$

$$f^{\infty}(y) = \sum_{i=1}^n e^{y_i - 1} \rightarrow y = -A^T \lambda - r \mathbf{1} \rightarrow f^{\infty}(-A^T \lambda - r \mathbf{1})$$

$$\sum_{i=1}^n e^{-a_i^T \lambda - r - 1}$$

dual $g(\lambda, r) = -\lambda^T b - r - \sum_{i=1}^n e^{-a_i^T \lambda - r - 1}$

$\max g(\lambda, r) \text{ \& } \lambda \geq 0 \rightarrow \sup (y^T x - \sum x_i \log x_i) \leftarrow \text{کد}$

$$\begin{aligned} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} - \begin{bmatrix} \log x_{i+1} \\ \vdots \\ \log x_{n+1} \end{bmatrix} &= 0 \rightarrow y_i = \log x_{i+1} \\ x_i &= e^{y_i - 1} \end{aligned}$$

$$f(y) = \sum_{i=1}^n y_i e^{y_i-1} - \sum e^{y_i-1} (y_i-1)$$

ادامه سوال (5)

$$= \sum e^{y_i-1} (y_i - y_{i+1}) = \sum_{i=1}^n e^{y_i-1}$$

$$\min_x f(x) = \|x\|^2$$

$$s.t. \quad c(x) = a^T x + \alpha \geq 0$$

$$L(x, \lambda) = f(x) - \lambda c(x)$$

سوال (6)

$$\|x\|^2 - \lambda (a^T x + \alpha)$$

$$\text{مثال: } \nabla_x L(x, \lambda) = 2x - \lambda a, \quad \nabla_{xx} L(x, \lambda) = 2I \quad ; \quad 2I \succ 0$$

$$\text{KKT: } \begin{cases} \nabla_x L(x^*, \lambda^*) = 0 \\ \lambda^* c(x^*) = 0 \\ \lambda^* \geq 0 \end{cases} \Rightarrow x^* = \frac{\lambda^*}{2} a$$

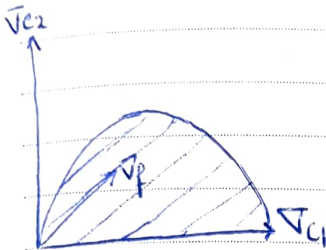
$$\lambda^* = 0 \quad \text{or} \quad a^T x^* + \alpha = \frac{\lambda^{*2} \|a\|^2}{2} + \alpha = 0$$

$$\text{if } \alpha \geq 0 \rightarrow \lambda^* = 0 \rightarrow (x^*, \lambda^*) = (0, 0)$$

$$\text{if } \alpha < 0 \rightarrow (x^*, \lambda^*) = \left(\frac{\alpha}{\|a\|^2} a, \frac{2}{\|a\|^2} \right)$$

$$\begin{aligned} \min \quad & x_1 + x_2 \\ \text{s.t.} \quad & x_1^2 + x_2^2 \leq 2 \\ & x_2 \geq 0 \end{aligned}$$

داده (7)



مجموعه مجاز: feasible

$$L(x, \lambda) = f(x) - \lambda_1 C_1(x) - \lambda_2 C_2(x)$$

$$\nabla_x L(x^*, \lambda^*) = 0, \lambda^* \geq 0$$

$$\lambda_1^* C_1(x^*) = 0, \lambda_2^* C_2(x^*) = 0$$

$$x^* = \begin{pmatrix} -\sqrt{2} \\ 0 \end{pmatrix} \rightarrow \nabla f(x^*) = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\nabla C_1(x^*) = \begin{pmatrix} 2\sqrt{2} \\ 0 \end{pmatrix}, \nabla C_2(x^*) = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \Rightarrow \lambda^* = \begin{pmatrix} \frac{1}{2\sqrt{2}} \\ 1 \end{pmatrix}$$