

Solution to Gate ST 2023 Q 26

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Question : Consider the following regression model

$$y_k = \alpha_0 + \alpha_1 \log_e k + \epsilon_k, \quad k = 1, 2, \dots, n,$$

where ϵ_k 's are independent and identically distributed random variables each having probability density function $f(x) = \frac{1}{2}e^{-|x|}, x \in \mathbb{R}$. Then which one of the following statements is true?

- (A) The maximum likelihood estimator of α_0 does not exist
- (B) The maximum likelihood estimator of α_1 does not exist
- (C) The least squares estimator of α_0 exists and is unique
- (D) The least squares estimator of α_1 exists, but it is not unique

Solution:

$$f(\epsilon_k) = \frac{1}{2}e^{-|\epsilon_k|} \quad (1)$$

$$\text{Likelihood function : } f(\epsilon_1 \epsilon_2 \dots \epsilon_n) = \prod_{k=1}^n f(\epsilon_k) \quad (2)$$

$$L = \prod_{k=1}^n \frac{1}{2}e^{-|\epsilon_k|} \quad (3)$$

$$L_1 = \ln L = \ln \left(\prod_{k=1}^n \frac{1}{2}e^{-|\epsilon_k|} \right) \quad (4)$$

$$= \sum_{k=1}^n \ln \left(\frac{1}{2}e^{-|\epsilon_k|} \right) \quad (5)$$

$$= \sum_{k=1}^n (-\ln 2 - |y_k - \alpha_0 - \alpha_1 \log_e k|) \quad (6)$$

$$= -n \ln 2 - \sum_{k=1}^n (|y_k - \alpha_0 - \alpha_1 \log_e k|) \quad (7)$$

$$L_1 = \text{function of } \alpha_0, \alpha_1 \quad (8)$$

1) Maximum likelihood estimator

We need to find the value of α_0 and α_1 which will maximise the value of L_1 i.e. the value of α_0 and

α_1 which will minimise the value of $\sum_{k=1}^n |y_k - \alpha_0 - \alpha_1 \log_e k|$

$$|y_k - \alpha_0 - \alpha_1 \log_e k| = \left| \mathbf{y}_{\mathbf{k}} - \begin{pmatrix} 1 & \log_e k \end{pmatrix} \begin{pmatrix} \alpha_0 \\ \alpha_1 \end{pmatrix} \right| \quad (9)$$

$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} \quad (10)$$

$$\mathbf{A} = \begin{pmatrix} 1 & \log_e 1 \\ 1 & \log_e 2 \\ \vdots & \vdots \\ 1 & \log_e n \end{pmatrix} \quad (11)$$

$$\mathbf{x} = \begin{pmatrix} \alpha_0 \\ \alpha_1 \end{pmatrix} \quad (12)$$

$$\begin{aligned} \min_{\alpha_0, \alpha_1} \quad & \mathbf{y} - \mathbf{Ax} \\ \text{s.t.} \quad & \mathbf{y} - \mathbf{Ax} \geq 0 \end{aligned}$$

$$\begin{aligned} \min_{\alpha_0, \alpha_1} \quad & -(\mathbf{y} - \mathbf{Ax}) \\ \text{s.t.} \quad & \mathbf{y} - \mathbf{Ax} \leq 0 \end{aligned}$$