

# EE23010 NCERT Exemplar

Vishal A - EE22BTECH11057

## Question 23.2023

Let  $X_1, X_2, X_3, \dots, X_n$  be a random sample of size  $n (\geq 2)$  from a population having probability density function

$$f(x; \theta) = \begin{cases} \frac{2}{\theta x} (\log_e x) e^{-\frac{(\log_e x)^2}{\theta}} & , 0 < x < 1 \\ 0 & , otherwise \end{cases}$$

where  $\theta > 0$  is an unknown parameter. The maximum likelihood estimator of  $\theta$  is,

**Solution:**

$$L(\theta) = f(x_1, x_2, \dots, x_n; \theta) \quad (1)$$

The product of pdfs can be used to approximate the likelihood function even if the variables are dependent. This is a general approach that is often used in practice to estimate MLE of  $\theta$ . Therefore,

$$L(\theta) = \prod_{i=1}^n f(x_i; \theta) \quad (2)$$

Maximizing  $L(\theta)$  is equivalent to maximizing the  $\ln L(\theta)$  as  $\ln$  is a monotonically increasing function.

$$l(\theta) = \ln L(\theta) \quad (3)$$

$$= \ln \left( \prod_{i=1}^n f(x_i; \theta) \right) \quad (4)$$

$$= \sum_{i=1}^n \ln f(x_i; \theta) \quad (5)$$

$$= -n \ln 2 - n \ln \theta + \sum_{i=1}^n \ln(-\ln x_i) - \sum_{i=1}^n (\ln x_i) - \sum_{i=1}^n \frac{(\ln x_i)^2}{\theta} \quad (6)$$

Maximizing  $l(\theta)$  with respect to  $\theta$  gives the MLE estimation, therefore

$$\frac{\partial l(\theta)}{\partial \theta} = 0 \quad (7)$$

$$\frac{-n}{\theta} + \frac{1}{(\theta)^2} \sum_{i=1}^n (\ln x_i)^2 = 0 \quad (8)$$

$$\theta = \frac{1}{n} \sum_{i=1}^n (\ln x_i)^2 \quad (9)$$