

## Question 39.2023

Anupama Kulshreshtha  
EE22BTECH11009

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

$$f(x; \mu) = \begin{cases} e^{-(x-\mu)}, & \text{if } \mu \leq x < \infty \\ 0, & \text{otherwise,} \end{cases} \quad (1)$$

where  $\mu \in \mathbb{R}$  is an unknown parameter. If  $\hat{M}$  is the maximum likelihood estimator of the median of  $X_1$ , then which one of the following statements is true?

A)  $P(\hat{M} \leq 2) = 1 - e^{-n(1-\log_e 2)}$  if  $\mu = 1$

B)  $P(\hat{M} \leq 1) = 1 - e^{-n \log_e 2}$  if  $\mu = 1$

C)  $P(\hat{M} \leq 3) = 1 - e^{-n(1-\log_e 2)}$  if  $\mu = 1$

D)  $P(\hat{M} \leq 4) = 1 - e^{-n(2 \log_e 2 - 1)}$  if  $\mu = 1$

**Solution:** For continuous random variable  $X$ , median  $M$  is such that,

$$P(X \leq M) = 0.5 \quad (2)$$

The pdf of  $X$  is given by,

$$p_X(x) = \begin{cases} e^{-(x-\mu)}, & \text{if } \mu \leq x < \infty \\ 0, & \text{otherwise,} \end{cases} \quad (3)$$

Hence, cdf is given by

$$F(x; \mu) = \int_{\mu}^x e^{-(t-\mu)} dt \quad (4)$$

$$= e^{\mu} [-e^{-x} + e^{-\mu}] \quad (5)$$

$$= 1 - e^{-(x-\mu)} \quad (6)$$

Now,

$$F(x; \mu) = 0.5 \quad (7)$$

$$\implies 1 - e^{-(M-\mu)} = 0.5 \quad (8)$$

$$\implies \hat{M} = \mu + \ln(2) \quad (9)$$

**Definition 1.**  $L$ , the Maximum Likelihood Estimator of the distribution is given by,

$$L = \prod e^{-(x-\mu)} \quad (10)$$

$$= e^{-(\sum x_i - n\mu)} \quad (11)$$

$$(12)$$

For the Likelihood function to be maximum,  $\sum x_i - n\mu$  should be minimum  
Hence,

$$X_i > \mu \quad (13)$$

$$\Rightarrow \sum x_i > n\mu \quad (14)$$

$$\sum x_i - n\mu > 0 \quad (15)$$

$$\Rightarrow \mu = \frac{\sum x_i}{n} \quad (16)$$

Given,

$$p_X(x) = e^{-(x-\mu)} \quad (17)$$

$$(18)$$

$X_i$  follows an exponential distribution.

**Definition 2.** We know that if,

$$p_X(x) = \lambda_i e^{-\lambda_i x} \quad (19)$$

$$S = X_1 + X_2 + \dots + X_n \quad (20)$$

$$p_S(n) = \frac{\lambda^n x^{n-1} e^{-\lambda x}}{(n-1)!} \quad (21)$$

which is gamma distribution with parameters  $n$  and  $\lambda$

Hence, pdf of

$$Y = \sum_{i=1}^n x_i \quad (22)$$

$$\text{where,} \quad (23)$$

$$p_X(x) = e^{-(x-\mu)} \quad (24)$$

will follow gamma distribution with parameter  $n$  and  $\lambda = 1$ , given by,

$$p_Y(x; n, 1) = \frac{x^{n-1} e^{-x}}{(n-1)!} \quad (25)$$

Hence, cdf is given by,

$$f_Y(x; n) = \int_1^x \frac{t^{n-1}e^{-t}}{(n-1)!} dt \quad (26)$$

$$= 1 - \Gamma(n, x) \quad (27)$$

where  $\Gamma(n, x)$  is incomplete gamma function. Thus,

$$P(\hat{M} \leq k) = P\left(\frac{\sum x_i}{n} + \ln(2) \leq k\right) \quad (28)$$

$$= P(Y/n + \ln(2) \leq k) \quad (29)$$

$$= P(Y \leq n(k - \ln(2))) \quad (30)$$

$$= f_Y(n(k - \ln(2))) \quad (31)$$

$$= 1 - \Gamma(n, n(k - \ln(2))) \quad (32)$$

It needs a value of n to be computed.