

Probability&RV Assignment-08

Anuradha U-ee21resch01008

Download Latex code from

https://github.com/Anuradha-Uggi/Assignments-AI5002-Probability-and-Random-Variables/blob/main/Prob_ass08/rvsp_8.tex

for all θ then

$$P_{\theta}(g(T) = 0) = 1 \quad (3)$$

for all θ .

I. QUESTION(UGC NET 2019,Q-108)

Suppose $X_i = X_1, X_2, \dots, X_n$ are i.i.d Uniform $(\theta, 2\theta)$, $\theta > 0$. Let $X_{(1)} = \min\{X_1, X_2, \dots, X_n\}$ and $X_{(n)} = \max\{X_1, \dots, X_n\}$. then which of the following statements are correct.

- 1) $(X_{(1)}, X_{(n)})$ is jointly sufficient and complete for θ
- 2) $(X_{(1)}, X_{(n)})$ is jointly sufficient but not complete for θ
- 3) $\frac{X_{(n)}}{2}$ is maximum likelihood estimate for θ
- 4) $X_{(1)}$ is maximum likelihood estimate for θ

II. BASIC DEFINITIONS

Sufficient Statistic:

Given X i.i.d Data conditioned on an unknown parameter θ , $T(X)$ is called sufficient statistic for θ if its values contains all the information needed to compute any estimate of the parameter (Maximum likelihood estimate). according to Fisher–Neyman Factorization PDF is

$$f(X; \theta) = h(X)S(\theta, T(X)) \quad (1)$$

where $h(X)$ is a constant and $S(\theta, T(X))$ is a function through which θ will interact to X only through $T(X)$.

Statistic Completeness:

$T(X)$ is said to be complete for θ if for every measurable function g ; if

$$E_{\theta}(g(T)) = 0 \quad (2)$$

III. SOLUTION

Given

$$X_{(1)} = \min\{X_1, X_2, \dots, X_n\} = \min\{X_i\} \quad (4)$$

$$X_{(n)} = \max\{X_1, X_2, \dots, X_n\} = \max\{X_i\} \quad (5)$$

$$X_i = \{X_1, X_2, \dots, X_n\} \sim U(\theta, 2\theta) \quad (6)$$

$$P(X_i) = \frac{1}{\theta} \quad (7)$$

Statistic

$$T(X) = (\min\{X\}, \max\{X\}) \quad (8)$$

X are i.i.d so Likelihood is given by

$$f_X(x) = \frac{1}{\theta^n} 1_{\{\theta \leq X_i \leq 2\theta\}} \quad (9)$$

equation (7) can be split into

$$f_X(x) = \frac{1}{\theta^n} 1_{\{\theta \leq \min\{X_i\}\}} 1_{\{\max\{X_i\} \leq 2\theta\}} \quad (10)$$

from equation (8)

$$h(X) = 1 \quad (11)$$

which is constant and

$$S_{(\theta, 2\theta)}(X) = \frac{1}{\theta^n} \quad (12)$$

which is function of only θ .

therefore $T(\min\{X_i\}, \max\{X_i\})$ is jointly sufficient to define θ thus Sufficient statistic.

Let

$$g(T) = \max\{X_i\} - \min\{X_i\} \quad (13)$$

$$E[g(T)] = E[\max\{X_i\} - \min\{X_i\}] = \int (\max\{X_i\} - \min\{X_i\}) \frac{1}{\theta^n} dx \quad (14)$$

from equation (12) its clear that

$$\max\{X_i\} - \min\{X_i\} \neq 0 \quad (15)$$

for all θ therefore

$$P(\max\{X_i\} - \min\{X_i\} = 0) \neq 1 \quad (16)$$

therefore $T(X_{(n)}, X_{(1)})$ is Jointly sufficient but not complete for θ .

Maximum Likelihood Estimate(MLE) :

Likelihood can be written as

$$f_{\theta}(X_1, X_2, \dots, X_n) = \frac{1}{\theta^n} I\left(\frac{\max\{X_i\}}{2} \leq \theta \leq \min\{X_i\}\right) \quad (17)$$

the MLE is the statistic that maximizes the likelihood. from equation (7) likelihood is a decreasing function of θ . therefore MLE of θ is

$$\theta = \frac{X_{(n)}}{2} = \frac{\max\{X_i\}}{2} \quad (18)$$

IV. CONCLUSION

From above observations option (2) and option (3) holds.