

CSIR UGC NET EXAM (Dec 2014), Question 113

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EP20BTECH11018

Some important concepts

- 1 The beta function, $B(x, y)$
- 2 Marginal distribution
- 3 Likelihood function and maximum likelihood estimator
- 4 Prior and posterior distributions
- 5 Posterior mean

The beta function, $B(x, y)$

The beta function, $B(x, y)$, is defined by the integral

$$\begin{aligned} B(x, y) &= \int_0^1 t^{x-1} (1-t)^{y-1} dt \\ &= \frac{x+y}{xy} \times \frac{1}{x+y C_x} \end{aligned} \quad (1)$$

where $\operatorname{Re}(x) > 0$ and $\operatorname{Re}(y) > 0$.

Marginal distribution

The marginal distribution of x is given by

$$f(x) = \int_y f(x, y) dy \quad (2)$$

Likelihood function

The likelihood function, represented by $f(X|\theta)$, refers to the joint probability of the data in the case of discrete distributions and joint probability density of the data in the case of continuous distributions, i.e.,

$$f(X|\theta) = \prod_{i=1}^n f(X_i|\theta) \quad (3)$$

Maximum Likelihood Estimator

The maximum likelihood estimator, represented by MLE, is the value which maximizes the likelihood function, i.e.,

$$\text{MLE} = \arg_{\theta} \max (f(X|\theta)) \quad (4)$$

Prior and Posterior Distribution

Prior distribution, where $f(\theta)$ is the prior density of θ , is the probability distribution that expresses established beliefs about an event before new evidence is taken into account.

When the new evidence is used to create a new distribution, that new distribution is called **posterior distribution**, where $f(\theta|X)$ is the posterior density of θ . From conditional probability we get,

$$f(\theta|X) = \frac{f(X, \theta)}{f(X)} \quad (5)$$

using (2) we get

$$f(\theta|X) = \frac{f(X, \theta)}{\int f(X, \theta) d\theta} \quad (6)$$

$$= \frac{f(X|\theta) f(\theta)}{\int f(X|\theta) f(\theta) d\theta} \quad (7)$$

Posterior mean

Posterior mean is the mean of the posterior distribution of θ , i.e.,

$$E(\theta|X) = \int \theta f(\theta|X) d\theta \quad (8)$$

Question

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Let X_1, X_2, \dots, X_n be independent and identically distributed Bernoulli(θ), where $0 < \theta < 1$ and $n > 1$. Let the prior density of θ be proportional to $\frac{1}{\sqrt{\theta(1-\theta)}}$, $0 < \theta < 1$. Define $S = \sum_{i=1}^n X_i$.

Then valid statements among the following are:

- ① The posterior mean of θ does not exist;
- ② The posterior mean of θ exists;
- ③ The posterior mean of θ exists and it is larger than the maximum likelihood estimator for all values of S .
- ④ The posterior mean of θ exists and it is larger than the maximum likelihood estimator for some values of S .

Solution

Let $f(\theta)$ be the prior density of θ .

$$\begin{aligned} f(\theta) &\propto \frac{1}{\sqrt{\theta(1-\theta)}} \\ \Rightarrow f(\theta) &= \frac{K}{\sqrt{\theta(1-\theta)}} \end{aligned} \quad (9)$$

where K is the proportionality constant.

$$\begin{aligned} \int_0^1 f(\theta) d\theta &= 1 \\ \Rightarrow K \int_0^1 \frac{1}{\sqrt{\theta(1-\theta)}} d\theta &= 1 \end{aligned} \quad (10)$$

Solution Contd.

From (1) we get,

$$K \times B\left(\frac{1}{2}, \frac{1}{2}\right) = 1$$
$$\Rightarrow K = \frac{1}{B\left(\frac{1}{2}, \frac{1}{2}\right)} \quad (11)$$

$$\therefore f(\theta) = \frac{\theta^{-\frac{1}{2}} (1 - \theta)^{-\frac{1}{2}}}{B\left(\frac{1}{2}, \frac{1}{2}\right)} \quad (12)$$

Solution Contd.

From (3) we get,

$$\begin{aligned} f(X|\theta) &= \prod_{i=1}^n \theta^{X_i} (1-\theta)^{1-X_i} \\ &= \theta^S (1-\theta)^{n-S} \end{aligned} \quad (13)$$

Using log of likelihood function and differentiating we get,

$$\ln(f(X|\theta)) = S \ln(\theta) + (n-S) \ln(1-\theta) \quad (14)$$

$$\frac{\partial \ln(f(X|\theta))}{\partial \theta} = \frac{S}{\theta} + \frac{S-n}{1-\theta} = 0$$

$$\therefore \text{MLE} = \frac{S}{n} \quad (15)$$

Solution Contd.

Using (12) and (13) in (7) we get,

$$f(\theta|X) = \frac{\theta^{S-\frac{1}{2}} (1-\theta)^{n-S-\frac{1}{2}}}{\int_0^1 \theta^{S-\frac{1}{2}} (1-\theta)^{n-S-\frac{1}{2}} d\theta} \quad (16)$$

Using (1) we get,

$$f(\theta|X) = \frac{\theta^{S-\frac{1}{2}} (1-\theta)^{n-S-\frac{1}{2}}}{B(S+\frac{1}{2}, n-S+\frac{1}{2})} \quad (17)$$

Solution Contd.

From (8) we get,

$$\begin{aligned} E(\theta|X) &= \int_0^1 \theta f(\theta|X) d\theta \\ &= \int_0^1 \frac{\theta^{S+\frac{1}{2}} (1-\theta)^{n-S-\frac{1}{2}}}{B(S+\frac{1}{2}, n-S+\frac{1}{2})} d\theta \\ &= \frac{B(S+\frac{3}{2}, n-S+\frac{1}{2})}{B(S+\frac{1}{2}, n-S+\frac{1}{2})} \end{aligned} \quad (18)$$

Using (1) we get

$$E(\theta|X) = \frac{S+\frac{1}{2}}{n+1} \quad (19)$$

Solution Contd.

For $E(\theta|X)$ to be greater than MLE,

$$\frac{S + \frac{1}{2}}{n + 1} > \frac{S}{n}$$
$$\therefore n > 2S \quad (20)$$

Solution Contd.

We see that

$$E(\theta|X) = \frac{S + \frac{1}{2}}{n + 1}$$

from (19) and

$$n > 1$$

given in the question $\implies E(\theta|X)$ exists.

From (20) we see that $E(\theta|X) > \text{MLE}$ for some values of S .

\therefore Options 2 and 4 are correct.