

Tyasha

102103034

3Cof2

Parameter Estimation



Ques 4.1 Random Sample = x_1, x_2, \dots, x_n from ND (Normal distribution) with mean θ_1 and Variance θ_2

Probability density funcⁿ is-

$$f(x, \theta_1, \theta_2) = \frac{1}{\sqrt{2\pi\theta_2}} e^{-\frac{(x-\theta_1)^2}{2\theta_2}} \quad \text{--- (i)}$$

Likelihood Function $L(\theta_1, \theta_2)$ is joint Pdf of sample

$$L(\theta_1, \theta_2) = \prod_{i=1}^n f(x_i, \theta_1, \theta_2) \quad \text{--- (ii)}$$

taking log of (ii), then (ii) becomes

$$\ln L(\theta_1, \theta_2) = -\frac{n}{2} \ln(2\pi) - \frac{n}{2} \ln(\theta_2) - \frac{1}{2\theta_2} \sum_{i=1}^n (x_i - \theta_1)^2$$

differentiate w.r.t θ_1 and Put = 0

$$\frac{\partial \ln L(\theta_1, \theta_2)}{\partial \theta_1} = -\frac{1}{\theta_2} \sum_{i=1}^n (x_i - \theta_1) = 0$$

$$\Rightarrow \sum_{i=1}^n (x_i - \theta_1) = 0$$

$$\sum_{i=1}^n x_i = n\theta_1$$

$$\left| \theta_1 = \frac{1}{n} \sum_{i=1}^n x_i \right|$$

So, MLE for θ_1 is Sample mean.

Now w.r.t θ_2 ,

$$\frac{\partial \ln L(\theta_1, \theta_2)}{\partial \theta_2} = -\frac{n}{2\theta_2} + \frac{1}{2\theta_2^2} \sum_{i=1}^n (x_i - \theta_1)^2 = 0$$

$$\Rightarrow -\frac{n}{2\theta_2} + \frac{1}{2\theta_2^2} \sum_{i=1}^n (x_i - \theta_1)^2 = 0$$

$$n = \frac{1}{\theta_2} \sum_{i=1}^n (x_i - \theta_1)^2$$

$$\left| \theta_2 = \frac{1}{n} \sum_{i=1}^n (x_i - \theta_1)^2 \right|$$

MLE for θ_2 is Sample Variance

Question 2
ay → Pdf of binomial distribution $B(n, p) = n C x p^x (1-p)^{n-x}$
 $n = m, p = \theta$
 $\therefore f(x_i) = m C x_i \theta^{x_i} (1-\theta)^{m-x_i}$

likelihood funⁿ $L(m, \theta) = \prod_{i=1}^n f(x_i) = \prod_{i=1}^n m C x_i \theta^{x_i} (1-\theta)^{m-x_i}$

$L(m, \theta) = \left(\prod_{i=1}^n m C x_i \right) \theta^{\sum_{i=1}^n x_i} (1-\theta)^{mn - \sum_{i=1}^n x_i}$

$\therefore \log L(m, \theta) = \log \left(\prod_{i=1}^n m C x_i \right) + \sum_{i=1}^n x_i \log \theta + (mn - \sum_{i=1}^n x_i) \log(1-\theta)$

$\frac{\partial \log L(m, \theta)}{\partial \theta} = \sum_{i=1}^n x_i \times \frac{1}{\theta} + \frac{1}{1-\theta} (mn - \sum_{i=1}^n x_i)$

$\frac{1}{1-\theta} (mn - \sum_{i=1}^n x_i) - \frac{1}{\theta} (\sum_{i=1}^n x_i) = 0$

$\frac{mn - \sum_{i=1}^n x_i}{\sum_{i=1}^n x_i} = \frac{1-\theta}{\theta}$
 $\therefore \theta_{MLE} = \frac{\sum x_i}{m}$