

PRIYAL SINGLA

102103274

3C010

Q1.

Normal distribution

$$\text{PMF}(x_i) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_i - \mu)^2}{2\sigma^2}}$$

$$\mu = \theta_1, \quad \sigma^2 = \theta_2$$

$$\therefore f(x_i | \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_i - \mu)^2}{2\sigma^2}}$$

$$f(x_i | \theta_1, \theta_2) = \frac{1}{\sqrt{2\pi\theta_2}} e^{-\frac{(x_i - \theta_1)^2}{2\theta_2}}$$

likelihood function,

$$L(\theta_1, \theta_2) = \prod_{i=1}^n f(x_i | \theta_1, \theta_2)$$

$$\Rightarrow L(\theta_1, \theta_2) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\theta_2}} e^{-\frac{(x_i - \theta_1)^2}{2\theta_2}}$$

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

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

$$\ln(L(\theta_1, \theta_2)) = \ln \left[(\theta_2)^{-n/2} (2\pi)^{-n/2} e^{-\frac{1}{2\theta_2} \sum_{i=1}^n (x_i - \theta_1)^2} \right]$$

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

differentiate w.r.t. θ_1 ,

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

Now, $\frac{\partial \ell(\theta_1, \theta_2)}{\partial \theta_1} = 0$

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

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

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

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

$$\Rightarrow \theta_1 = \bar{x}_n$$

$$\therefore \boxed{\theta_{\text{MLE}} = \bar{x}_n} \quad \text{--- (2)}$$

differentiating ① w.r.t ②,

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

now $\frac{\partial \ell(\theta_1, \theta_2)}{\partial \theta_2} = 0$

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

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

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

from equation ②,

$$\boxed{\theta_{2MLE} = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x}_n)^2}$$

Q2:

for binomial distribution,

$$PMF(x_i) = {}^n C_{x_i} p^{x_i} (1-p)^{n-x_i}$$

$$n = m, \quad p = \theta$$

$$L(p) = \prod_{i=1}^n P(x_i | m, \theta)$$

$$= \prod_{i=1}^n \left({}^m C_{x_i} \theta^{x_i} (1-\theta)^{m-x_i} \right)$$

$$= \prod_{i=1}^n {}^m C_{x_i} \cdot \prod_{i=1}^n \theta^{x_i} \cdot \prod_{i=1}^n (1-\theta)^{m-x_i}$$

$$= \prod_{i=1}^n {}^m C_{x_i} \cdot \theta^{\sum_{i=1}^n x_i} (1-\theta)^{nm - \sum_{i=1}^n x_i}$$

$$\ln L(p) = \ln \left(\prod_{i=1}^n {}^m C_{x_i} \cdot \prod_{i=1}^n \theta^{x_i} (1-\theta)^{nm - \sum_{i=1}^n x_i} \right)$$

$$= \ln \left(\prod_{i=1}^n {}^m C_{x_i} \right) + \ln \left(\theta^{\sum_{i=1}^n x_i} \right) + \ln \left((1-\theta)^{nm - \sum_{i=1}^n x_i} \right)$$

$$= \ln \left(\prod_{i=1}^n {}^m C_{x_i} \right) + \ln(\theta) \cdot \sum_{i=1}^n x_i + \ln(1-\theta) \left(nm - \sum_{i=1}^n x_i \right)$$

diff. w.r.t θ

$$\frac{\partial L(P)}{\partial \theta} = \frac{1}{\theta} \sum_{i=1}^n x_i + \left(\frac{-1}{1-\theta} \right) \left(nm - \sum_{i=1}^n x_i \right)$$

$$= \frac{1}{\theta} \sum_{i=1}^n x_i - \frac{1}{1-\theta} \left(nm - \sum_{i=1}^n x_i \right)$$

now $\frac{\partial L(P)}{\partial \theta} = 0$

$$\therefore \frac{1}{\theta} \sum_{i=1}^n x_i - \frac{1}{1-\theta} \left(nm - \sum_{i=1}^n x_i \right) = 0$$

$$\Rightarrow \frac{1}{\theta} \sum_{i=1}^n x_i = \frac{1}{1-\theta} \left(nm - \sum_{i=1}^n x_i \right)$$

$$\frac{1-\theta}{\theta} = \frac{nm - \sum_{i=1}^n x_i}{\sum_{i=1}^n x_i}$$

$$\Rightarrow \frac{nm}{\sum_{i=1}^n x_i} - 1 = \frac{1}{\theta} - 1$$

$$\Rightarrow \theta = \frac{\sum_{i=1}^n x_i}{nm}$$

$$\Rightarrow \theta = \frac{\bar{x}_n}{m}$$

$$\therefore \theta_{MLE} \in (0,1) = \frac{\bar{x}_n}{m}$$