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Parameter Estimation Assignment

Q1 let (x_1, x_2, \dots) be a random sample of size n taken from a normal population with parameters mean $= \theta_1$ and variance $= \theta_2$. Find the maximum likelihood estimate of these two parameters

Sol
$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad (\because \text{PDF of Normal distribution})$$

$$\mu_1 = \theta_1, \sigma^2 = \theta_2$$

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

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

Likelihood Function

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

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

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

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

Taking log Both Sides

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

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

Diff eqn (2) w.r.t θ_1

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

now,

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

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

$$\Rightarrow n\theta_1 = \sum_{i=1}^n x_i \Rightarrow \theta_1 = \frac{\sum_{i=1}^n x_i}{n} \Rightarrow \theta_{1MLE} = \bar{x}_n \quad \text{--- (3)}$$

Diff (2) w.r.t θ_2

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

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

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

From (3) $\theta_1 = \bar{x}_n$

$$\theta_2 = \frac{1}{n} \sum_{i=1}^n (x_i^o - \bar{x}_n)^2$$

Ques 2 Let x_1, x_2, \dots, x_n be random sample from $B(m, \theta)$ distribution where $\theta \in \mathcal{D} = (0, 1)$ is unknown and 'm' is a known true integer. Compute value of θ using MLE

Sol: $f(n) = {}^n C_x p^n (1-p)^{n-x}$ (PDF of Binomial Distribution)

here, $n = m$ & $p = \theta$

$$b(x) = {}^m C_x \theta^x (1-\theta)^{m-x} \Rightarrow f(x_i^o) = {}^m C_{x_i^o} \theta^{x_i^o} (1-\theta)^{m-x_i^o}$$

Likelihood function

$$L(m, \theta) = \prod_{i=1}^n f(x_i^o)$$

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

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

Taking log Both sides.

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

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

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

Diff ① w.r.t θ

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

Now,

$$\frac{\partial Z}{\partial \theta} = 0$$

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

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

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

$$+ \frac{mn}{\sum x_i} = 1 - \frac{1}{\theta}$$

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

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