

Assignment: Parameter Estimation

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Section: 3C010

Q1) density function for normal distribution with mean μ & variance σ^2

$$pmf(x_i) = \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right) \times e^{-\frac{(x_i - \mu)^2}{2\sigma^2}}$$

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

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

$$f(x_i | \theta_1, \theta_2) = \frac{1}{\sqrt{2\pi\theta_2}} e^{-\frac{(x_i - \theta_1)^2}{2\theta_2}} \quad \begin{matrix} \theta_1 = \mu \\ \theta_2 = \sigma^2 \end{matrix}$$

likelihood function

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

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

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

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

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

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

Differentiate w.r.t θ_1

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

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

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

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

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

$\therefore \theta_1 = \bar{x}_n \rightarrow (2)$

MLE

Differentiating (1) w.r.t (2)

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

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

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

$$\frac{1}{2\sigma_2^2} \sum_{i=1}^n (x_i - \sigma_1)^2 = -\frac{n}{2\sigma_2^2}$$

$$\boxed{\sigma_2^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \sigma_1)^2}$$

from (2)

$$\boxed{\sigma_2^2 \text{ MLE} = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x}_n)^2}$$

Q2)

For Binomial Distribution, density function is :-

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

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

$$\begin{aligned} L(p) &= \prod_{i=1}^n P(x_i | m, \theta) \\ &= \prod_{i=1}^n ({}^m C_{x_i} \cdot \theta^{x_i} (1-\theta)^{m-x_i}) \end{aligned}$$

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

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

$$\begin{aligned} \ln(L(p)) &= \ln \left[\prod_{i=1}^n {}^m C_{x_i} \theta^{\sum_{i=1}^n x_i} (1-\theta)^{nm - \sum_{i=1}^n x_i} \right] \\ &= \ln \left(\prod_{i=1}^n {}^m C_{x_i} \right) + \ln \left(\theta^{\sum_{i=1}^n x_i} \right) + \ln \left((1-\theta)^{nm - \sum_{i=1}^n x_i} \right) \quad (\text{P.T.O}) \end{aligned}$$

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

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

diff. wrt θ .

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

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

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

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

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

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

$$\theta = \frac{\sum_{i=1}^n x_i}{nm} = \frac{\bar{x}_n}{nm}$$

$$\theta_{MLE} = \frac{\bar{x}_n}{nm}$$