

Parameter Estimation Assignment.

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Ques 1. Let (x_1, x_2, \dots) be a random sample of size n taken from a normal population with parameters: mean = θ_1 and variance = θ_2 . Find the Maximum Likelihood Estimates of these two parameters.

Ans: A sample of size 'n' is taken: (x_1, x_2, \dots, x_n)

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

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

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

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

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

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

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

Taking log.

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

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

$$\text{Loss} = -\frac{n}{2} \ln(\theta_2) - \frac{n}{2} \ln 2\pi - \frac{1}{2\theta_2} \left\{ \sum_{i=1}^n (x_i - \theta_1)^2 \right\} - 0$$

Dif. both sides w.r.t. θ_1

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

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

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

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

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

$$\theta_1 \text{ me} = \bar{x}_n \rightarrow ②$$

Dif. w.r.t. θ_2

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

Putting equals to 0.

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

$$\Rightarrow \frac{1}{\hat{\theta}_2(\hat{\theta}_2)^2} \sum_{i=1}^n (x_i^o - \hat{\theta}_1)^2 = -\frac{n}{2\hat{\theta}_2}$$

$$\hat{\theta}_2 = \frac{1}{n} \sum_{i=1}^n (x_i^o - \hat{\theta}_1)^2 - \textcircled{3}$$

from \textcircled{2} & \textcircled{3}

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

Ques Let x_1, x_2, \dots, x_n be a random sample from $B(m, \theta)$ distribution where $\theta \in (0, 1)$ is unknown and m is a known positive integer. Compute values of θ using the M.L.E.

Soln For binomial distribution,

$$PMF = {}^n C_x p^x (1-p)^{n-x}$$

$$\text{Given: } n = m \quad p = \theta.$$

$$= {}^m C_x \theta^x (1-\theta)^{m-x}$$

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

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

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

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

Taking log

$$\ln(L(p)) = \ln\left(\prod_{l=1}^n \theta^{x_l^o} (1-\theta)^{nm - \sum_{l=1}^n x_l^o}\right)$$

$$= \ln\left(\prod_{l=1}^n \theta^{x_l^o}\right) + \ln\left(\theta^{\sum_{l=1}^n x_l^o}\right) + \ln((1-\theta)^{nm - \sum_{l=1}^n x_l^o})$$

$\sum_{l=1}^n x_l^o \ln(\theta) \cancel{\neq 0}$

- Diff. w.r.t. θ

$$\frac{\partial \ln(L(p))}{\partial \theta} = \frac{1}{\theta} \sum_{l=1}^n x_l^o - \frac{1}{1-\theta} (nm - \sum_{l=1}^n x_l^o) = 0$$

$$= \frac{1}{\theta} \sum_{l=1}^n x_l^o = \frac{1}{1-\theta} (nm - \sum_{l=1}^n x_l^o)$$

$$\frac{1-\theta}{\theta} = \frac{nm - \sum_{l=1}^n x_l^o}{\sum_{l=1}^n x_l^o}$$

$$\frac{nm}{\sum_{l=1}^n x_l^o} - 1 = \frac{1}{\theta} - 1$$

$$\theta = \frac{\sum_{l=1}^n x_l^o}{nm}$$

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

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