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Page No.

Date: / / 20

Parametric Estimation Assignment

Ques 1

$(x_1, x_2, \dots, x_n) \rightarrow$ random sample

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

not a ref. to the 2

$\mu \rightarrow$ mean

$\sigma \rightarrow$ variance

also not a 2 point

MLE of 2 parameters

42

$$L(x_1, x_2, \dots, x_n) = \prod_{i=1}^n L(x_i) = \prod_{i=1}^n \left(\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_i-\mu)^2}{2\sigma^2}} \right)$$

$$= \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_1-\mu)^2}{2\sigma^2}} \times \dots \times \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_n-\mu)^2}{2\sigma^2}}$$

$$L(x_1, x_2, \dots, x_n) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_1-\mu)^2}{2\sigma^2}} \times \dots \times \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_n-\mu)^2}{2\sigma^2}}$$

taking log on both sides

$$\ln\left(\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_1-\mu)^2}{2\sigma^2}}\right) = \ln\left(\frac{1}{\sqrt{2\pi\sigma^2}}\right) + \ln\left(e^{-\frac{(x_1-\mu)^2}{2\sigma^2}}\right)$$

$$\ln\left(\frac{1}{\sqrt{2\pi\sigma^2}}\right) = \ln(2\pi\sigma^2)^{-\frac{1}{2}} \dots (2)$$

(2) in (1) and $\sigma = \sigma^2$

$$= \frac{1}{2} \ln(2\pi\sigma^2) - \frac{(x_1-\mu)^2}{2\sigma^2} \ln(\sigma^2)$$

$$\approx \frac{-1}{2} (\ln 2\pi + \ln \theta) - \frac{(x_1 - \mu)^2}{2\theta}$$

$$\approx \frac{-1}{2} \ln 2\pi - \frac{1}{2\theta} \ln \theta - \frac{(x - \mu)^2}{2\theta} = \ell(x)$$

similar for n -terms

taking $\frac{\delta}{\delta \mu}$ on both sides.

$$\frac{\delta}{\delta \mu} \ell(x_1, \dots, x_n) = 0 = \frac{\delta}{\delta \mu} \left(-\frac{n}{2} \ln \theta - \frac{1}{2\theta} \sum_{i=1}^n (x_i - \mu)^2 \right)$$

$$= \frac{1}{\theta} (x_1 - \mu) \dots (x_n - \mu)$$

$$= \frac{1}{\theta} [(x_1 + x_2 + \dots + x_n) - n\mu] \quad \dots (2)$$

Now $\frac{\delta}{\delta \theta}$

$$\frac{\delta}{\delta \theta} \ell(x_1, \dots, x_n) = 0 = -\frac{n}{2\theta} + \frac{(x_1 - \mu)^2}{\theta^2} + \dots + \frac{(x_n - \mu)^2}{\theta^2}$$

$$= -\frac{n}{2\theta} + \frac{1}{\theta^2} ((x_1 - \mu)^2 + \dots + (x_n - \mu)^2)$$

MLE over $\mu = 0$

$$(2) = 0$$

$$0 = \frac{1}{n} \left[\sum_{i=1}^n (x_i - \mu) \right]$$

$$x_1 + x_2 + \dots + x_n = n\mu$$

$$\mu = \frac{x_1 + x_2 + \dots + x_n}{n} \rightarrow \text{mean}$$

MLE under $\theta = 0$
 (x_1, \dots, x_n)
 $\theta = 0$

$$0 = -\frac{n}{2\sigma^2} + \frac{1}{2\sigma^2} \left[\sum_{i=1}^n (x_i - \mu) + \dots + (x_n - \mu) \right]$$

$$0 = \frac{1}{2\sigma^2} \left[\sum_{i=1}^n (x_i - \mu) \right]$$

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variance
 σ^2

$$0 = \frac{1}{2\sigma^2} \left[\sum_{i=1}^n (x_i - \mu) \right]$$

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Ques 2

Binomial distribution \rightarrow
 $x_1, \dots, x_n \in B(m, \theta)$

$$\theta \in \Theta(0, 1)^+$$

m is the

find θ

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

$$L(x_1, \dots, x_n) = \prod_{i=1}^n P(x_i, \theta)$$

taking log on both sides.

$$\ln(L(x_1, \dots, x_n)) = \sum_{i=1}^n \ln({}^m C_{x_i}) + x_i \ln \theta + (m - x_i) \ln(1-\theta)$$

Now $\frac{\partial}{\partial \theta}$ on both sides

so

$$\frac{\partial}{\partial \theta} \ln L(x_1, \dots, x_n) = 0 + \sum_{i=1}^n \frac{x_i - m + x_i}{\theta(1-\theta)}$$

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

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

$$MCE \text{ over } \theta = 0$$

$$(1) = 0$$

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

$$x_i - m\theta = 0$$

$$\theta = \frac{\sum_{i=1}^n x_i}{mn} = \frac{\bar{x}}{m}$$

$$\left(\sum_{i=1}^n \frac{x_i}{n} = \bar{x} \right) \quad \downarrow \text{mean}$$

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

$$\therefore \bar{x} = m\theta$$

$$\text{or}$$