

# # Predictive Analytics using Statistics : (UCS 654)

## # Assignment -> Parameter Estimation.....

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Q. Given  $(x_1, x_2, \dots)$  -> random sample of size  $n$  (from normal population)

• mean =  $\theta_1$ , variance =  $\theta_2$

• To find max likelihood estimates of these two parameters.

Soln) The likelihood function  $L(\theta_1, \theta_2)$  is given by :

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

To maximize this function, we will take natural logarithm of  $L(\theta_1, \theta_2)$

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

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

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

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

Now we need to maximise likelihood w.r.t  $\theta_1$  &  $\theta_2$  separately.



• w.r.t  $\theta_1$

• Differentiate  $l(\theta_1, \theta_2)$  w.r.t  $\theta_1$  & equate with 0  
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$$\frac{\partial l}{\partial \theta_1} = \sum_{i=1}^n \frac{1}{2\theta_2} (x_i - \theta_1) (-1) = 0$$

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

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

$$\theta_1 = \frac{\left( \sum_{i=1}^n x_i \right)}{n} \quad (\text{max likelihood for } \theta_1 \text{ is sample mean})$$

• w.r.t  $\theta_2$

• Differentiate  $l(\theta_1, \theta_2)$  w.r.t  $\theta_2$  & equate with 0

$$\frac{\partial l}{\partial \theta_2} = \sum_{i=1}^n \left( \frac{-1}{2\theta_2^2} x_i + \frac{(x_i - \theta_1)^2}{2\theta_2^3} \right) = 0$$

$$\Rightarrow \sum_{i=1}^n \left( \frac{-1}{2\theta_2^2} + \frac{(x_i - \theta_1)^2}{2\theta_2^3} \right) = 0$$

$$\sum_{i=1}^n \frac{(x_i - \theta_1)^2}{2\theta_2^3} = \frac{n}{2\theta_2^2}$$

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

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

Thus, max likelihood estimates are:

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

(mean)  
•  $n \rightarrow$  unbiased

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

(sample variance)  
estimator

2) Given,  $X_1, X_2, \dots, X_n \rightarrow$  random sample from Binomial dist.  $(m, \theta)$ .

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$\rightarrow \theta \in \Theta = (0, 1)$ ,  $m > 1$ ve integer.  
To find  $\theta$  using MLE

(soln) For binomial dist.  $B(m, \theta)$ , likelihood func. is given by  $\rightarrow$

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

Taking logarithm of the likelihood function  $\rightarrow$

$$\ell(\theta) = \sum_{i=1}^n \left[ \ln \binom{m}{x_i} + x_i \ln(\theta) + (m-x_i) \ln(1-\theta) \right]$$

To find ~~max~~ MLE of  $\theta$ , we take derivative of log likelihood w.r.t  $\theta$  & equate with 0

$$\frac{d\ell}{d\theta} = \sum_{i=1}^n \left[ \frac{x_i}{\theta} + \frac{(m-x_i)}{(1-\theta)} \times -1 \right]$$

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

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

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

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



$$\Rightarrow \sum_{i=1}^n x_i - \theta \sum_{i=1}^n \frac{x_i}{m} = 0; n \cdot m - \theta \sum_{i=1}^n \frac{x_i}{m}$$

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$$\Rightarrow 0 = \frac{\sum x_i}{nm} \Rightarrow \frac{\bar{x}}{m}$$

$\therefore$  MLE of  $\theta$  is  $\frac{\bar{x}}{m}$ , & sample mean