

Chapter 9 Estimation problems

- population parameter
 - Point estimation
 - Interval estimation

Another way is to approximate using Maximum likelihood estimation

- Binomial process \rightarrow p is the parameter associated
You perform $\textcircled{100}$ trials & you get a sample

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On the basis of this sample you will try to find p .

- Suppose my binomial experiment 3 trials
 trial $\begin{cases} D (p) \\ N (1-p) \end{cases}$ (we want to find out p)

When I actually performed the experiment, suppose I get

$\begin{cases} p \\ 1-p \end{cases}$ \rightarrow $\textcircled{D} D N$

probability of getting this outcome = $\underbrace{p \cdot p \cdot (1-p)}_{\text{maximize it}}$ Likelihood function

MLE gives you a point estimate

$$L(p) = p^2(1-p)$$

$$L'(p) = 0 \Rightarrow 2p - 3p^2 = 0$$

$$\Rightarrow p = 0 \text{ or } \textcircled{p = \frac{2}{3}} \text{ Ans.}$$

\downarrow
estimation for p by MLE

Formally MLE (maximum likelihood estimation)

Given independent observations x_1, x_2, \dots, x_n from a pdf / pmf

$$L(x_1, x_2, \dots, x_n, \theta) = \underbrace{f(x_1, \theta)}_{\checkmark} \cdot \underbrace{f(x_2, \theta)}_{\checkmark} \cdot \dots \cdot \underbrace{f(x_n, \theta)}_{\checkmark}$$

\downarrow
Likelihood function
maximize this function by using $L'(\theta) = 0$, that gives you an estimation for θ .

Que Consider samples $(x_1) x_2 \dots x_n$ from a normal distribution $N(\mu, \sigma)$. Find the maximum likelihood estimators of μ & σ .

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

Soln Likelihood function

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

$$L = \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2}\left(\frac{x_i-\mu}{\sigma}\right)^2\right)$$

$$L = \frac{1}{(\sqrt{2\pi})^n \sigma^n} \exp\left(-\frac{1}{2} \sum_{i=1}^n \left(\frac{x_i-\mu}{\sigma}\right)^2\right)$$

$$\ln L = \underbrace{\left(\frac{n}{2} \ln(2\pi)\right)}_{\text{constant}} - \underbrace{\left(\frac{n}{2} \ln \sigma^2\right)}_{\text{constant}} - \frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2$$

$$\ln L = \underbrace{\left(\frac{n}{2} \ln(2\pi) \right)}_{(n, \sigma)} - \underbrace{\left(\frac{n}{2} \ln \sigma^2 \right)}_{\sigma^2} - \frac{1}{2} \sum_{i=1}^n \left(\frac{x_i - \mu}{\sigma} \right)^2$$

differentiate with respect to μ

$$\frac{\partial \ln L}{\partial \mu} = -\frac{1}{\sigma} \sum_{i=1}^n \left(\frac{x_i - \mu}{\sigma} \right) (-1) = 0$$

For σ^2
differentiate w.r.t σ^2

$$\frac{\partial \ln L}{\partial \sigma^2} = -\frac{n}{2\sigma^2} + \frac{1}{2\sigma^4} \sum_{i=1}^n (x_i - \mu)^2$$

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

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

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

MLE estimator of μ .

$$\frac{n}{2\sigma^2} = \frac{1}{2\sigma^4} \sum_{i=1}^n (x_i - \mu)^2$$

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

MLE estimator for σ^2 is not unbiased.

Que Suppose 10 rats are used in a biomedical study where they are injected with cancer cells and then given a cancer drug that is designed to increase their survival rate.

The survival times, in months, are 14, 17, 27, 18, 12, 8, 22, 13, 19 & 12. Assume that the exponential distribution applies.

Give a maximum likelihood estimate of the mean survival rate.

Soln

exponential distribution

$$f(x, \beta) = \begin{cases} \frac{1}{\beta} e^{-x/\beta} & x > 0 \\ 0 & \text{otherwise} \end{cases}$$

$\beta \rightarrow$ mean value.

$$L(\beta) = \left(\frac{1}{\beta} e^{-\frac{x_1}{\beta}} \right) \left(\frac{1}{\beta} e^{-\frac{x_2}{\beta}} \right) \dots \left(\frac{1}{\beta} e^{-\frac{x_{10}}{\beta}} \right)$$

$$L(\beta) = \frac{1}{\beta^{10}} e^{-\frac{1}{\beta} \sum_{i=1}^{10} x_i} \rightarrow \ln L = -10 \ln \beta - \frac{1}{\beta} \sum_{i=1}^{10} x_i$$

$$\frac{\partial L}{\partial \beta} = -\frac{10}{\beta} + \frac{1}{\beta^2} \sum_{i=1}^{10} x_i = 0 \Rightarrow \beta = \frac{1}{10} \sum_{i=1}^{10} x_i = 16.2$$

$$\frac{d^2 L}{d\beta^2} = \frac{10}{\beta^2} - \frac{2}{\beta^3} \sum_{i=1}^{10} x_i$$

this no. is -ve for $\beta = 16.2$

H.W

Que

It is known that a sample consisting of values 12, 11.2, 13.5, 12.3, 13.8, & 11.9 comes from a population with pdf

$$f(x, \theta) = \begin{cases} \frac{\theta}{x^{\theta+1}} & x > 1 \\ 0 & \text{otherwise} \end{cases}$$

where $\theta > 0$.

$$f(x, \theta) = \begin{cases} \frac{\theta}{x^{\theta+1}} & x > 1 \\ 0 & \text{otherwise} \end{cases} \quad \text{where } \theta > 0.$$

Find the MLE of θ .

Solⁿ Ans is 0.3970 (Do it yourself).

End of chapter α

Chapter 10 (last chapter)

Test of Hypothesis

- You make a claim \leftarrow Hypothesis
- A statistical procedure to reject or fail to reject the hypothesis.

Hypothesis \div A person is guilty.

Steps for Hypothesis testing:-

Step I We will always state the given Hypothesis in the following way:-

Null Hypothesis H_0 : $\mu = 68 \rightarrow$ what you are claiming.

Claim population mean $\mu = 68$

Alternative " H_1 : $\mu \neq 68 \rightarrow$ contraction to null hypothesis.

4 more steps \leftarrow

Decision Step $\left\{ \begin{array}{l} \text{We reject } H_0 \text{ in favor of } H_1, \\ \text{We fail to reject } H_0. \end{array} \right\}$ These will be our final statements.