

Statistical inference

lecture3

Chapter Two: Estimation

Properties of good estimator

Lecturer

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1-2-3 Consistency

Consistency

Def let X_1, \dots, X_n be n.v from a population with p.d.f, $f(x, \theta)$, $\theta \in \mathcal{R}$

The estimator $\hat{\theta}_n$ is called consistency estimator for θ iff:

$$\lim_{n \rightarrow \infty} E(\hat{\theta} - \theta) = 0$$

where

$$1) \lim_{n \rightarrow \infty} \text{Var}(\hat{\theta}) = 0$$

$$2) \lim_{n \rightarrow \infty} \beta^2 = 0$$

$$3) \lim_{n \rightarrow \infty} \text{MSE}(\hat{\theta}) = 0$$

Example(1)

let $x \sim \beta(n, \theta)$

is \bar{x}, x_i consis. est. for θ ?

sol

$$\begin{aligned} \textcircled{1} \lim_{n \rightarrow \infty} \text{MSE}(\bar{x}) &= \lim_{n \rightarrow \infty} \frac{\theta(1-\theta)}{n} \\ &= \theta(1-\theta) \lim_{n \rightarrow \infty} \frac{1}{n} \end{aligned}$$

note:
 $\frac{\theta(1-\theta)}{n}$ size

$$\therefore \lim_{n \rightarrow \infty} \text{MSE}(\bar{x}) = 0$$

$\therefore \bar{x}$ is consis. est. for θ

(1)

$$\textcircled{2} \lim_{n \rightarrow \infty} \text{MSE}(X_1) = \lim_{n \rightarrow \infty} \theta(1-\theta) \\ = \theta(1-\theta)$$

$\therefore X_1$ is not consis. est. for θ

Example(2) IF $X \sim N(\mu, \sigma^2)$ is ar.v of size (n) , i) is \bar{X} consis. est. for μ ?
ii) is S^2 consis. est. for σ^2 ?

Sol

$$i) \therefore E(\bar{X}) = \mu, \text{Var}(\bar{X}) = \frac{\sigma^2}{n}$$

$$\lim_{n \rightarrow \infty} \text{var}(\bar{X}) = \lim_{n \rightarrow \infty} \frac{\sigma^2}{n} = 0$$

$$\lim_{n \rightarrow \infty} B^2(\bar{X}) = \lim_{n \rightarrow \infty} (0)^2 = 0$$

$\therefore \bar{X}$ is a consis. est. for μ

$$ii) S^2 = \frac{\sum (X_i - \bar{X})^2}{n}$$

$$\therefore E(S^2) = \frac{n-1}{n} \sigma^2$$

$$B^2 = [E(S^2) - \sigma^2]^2$$

$$= \left[\sigma^2 \frac{n-1}{n} - \sigma^2 \right]^2$$

$$= \left[\cancel{\sigma^2} - \frac{\sigma^2}{n} - \cancel{\sigma^2} \right]^2$$

$$B^2 = \frac{\sigma^4}{n^2}$$

$$\therefore \lim_{n \rightarrow \infty} \beta^2 = \lim_{n \rightarrow \infty} \frac{\sigma^4}{n^2} = 0$$

$$\text{Var}(S^2) = ?$$

$$\frac{nS^2}{\sigma^2} = \frac{n}{\sigma^2} \cdot \frac{\sum (x_i - \bar{x})^2}{n} \sim \chi^2_{(n-1)}$$

χ^2 $\left\{ \begin{array}{l} E(x) = m \\ \text{Var}(x) = 2m \end{array} \right.$

$$\therefore \text{Var}\left(\frac{nS^2}{\sigma^2}\right) = 2(n-1)$$

$$\therefore \text{Var}(S^2) = \frac{2(n-1)}{n^2} \sigma^4$$

$$\therefore \lim_{n \rightarrow \infty} \text{Var}(S^2) = \lim_{n \rightarrow \infty} \frac{2(n-1)}{n^2} \sigma^4$$

$$= \sigma^4 \lim_{n \rightarrow \infty} \left(\frac{2n}{n^2} - \frac{2}{n} \right)$$

$$= 0$$

$\therefore S^2$ is consis. est. for σ^2

Note IF $X \sim N(\mu, \sigma^2)$

$$Z = \frac{X - \mu}{\sigma} \sim N(0, 1)$$

$$\left(\frac{X - \mu}{\sigma}\right)^2 \sim \chi^2_{(1)}$$

$$\sum \left(\frac{X - \mu}{\sigma}\right)^2 \sim \chi^2_{(n)}$$

$$\frac{\sum (x_i - \bar{x})^2}{\sigma^2} + \frac{n(\bar{x} - \mu)^2}{\sigma^2} \sim \chi^2_{(n)}$$

$$\chi^2_{(n-1)} + \chi^2_{(1)} \sim \chi^2_{(n)}$$

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