

LECTURE - (3)

Agenda:

Examples to understand unbiasedness, MSE

EXAMPLE 1 (Exercise 8.10)

The number of breakdowns per week for a type of minicomputer is a random variable Y with a Poisson distribution with mean λ . A "random sample" Y_1, Y_2, \dots, Y_n of observations on the weekly number of breakdowns is available.

(a) Suggest an unbiased estimator for λ .

IMPORTANT: IN STANDARD STATISTICAL TERMINOLOGY,

UNLESS OTHERWISE STATED, A "RANDOM SAMPLE

Y_1, Y_2, \dots, Y_n FROM A POPULATION" MEANS THAT

Y_1, Y_2, \dots, Y_n ARE INDEPENDENT AND HAVE ~~THE~~

~~THE~~ THE SAME DISTRIBUTION.

In this case, it means that Y_1, Y_2, \dots, Y_n are independent and have a Poisson distribution with mean λ .

Note that

$$E[Y_i] = \lambda \quad \text{for each } i = 1, 2, \dots, n.$$

Hence, the most obvious unbiased estimator of λ is

$$\hat{\lambda}_{\text{avg}} = \frac{1}{n} \sum_{i=1}^n Y_i = \bar{Y}.$$

In statistics, the average of random variables Y_1, Y_2, \dots, Y_n is often represented by \bar{Y} .

Note that

$$E[\hat{\lambda}_{\text{avg}}] = E[\bar{Y}] = \frac{1}{n} \sum_{i=1}^n E[Y_i] = \frac{1}{n} \sum_{i=1}^n \lambda = \lambda.$$

Hence $\hat{\lambda}_{\text{avg}}$ is an unbiased estimate of λ .

However, one may just $\hat{\lambda}_1 = Y_1$ as an estimator for λ .

Note that

$$E[\hat{\lambda}_1] = E[Y_1] = \lambda.$$

Hence $\hat{\lambda}_1$ is also an unbiased estimator for λ . So why would we not work with $\hat{\lambda}_1$. The reason is simple. Choose the estimator with the smaller MSE.

Since $\hat{\lambda}_{avg}$ and $\hat{\lambda}_d$ are both unbiased,

$$MSE(\hat{\lambda}_{avg}) = \text{Var}(\hat{\lambda}_{avg}) = \frac{1}{n^2} V(\sum_{i=1}^n Y_i) = \frac{\lambda}{n}$$

$$MSE(\hat{\lambda}_d) = \text{Var}(\hat{\lambda}_d) = V(Y_d) = \lambda$$

Hence, $\hat{\lambda}_{avg}$ ~~is~~ always has a lower MSE than $\hat{\lambda}_d$.

(b) The weekly cost of repairing these breakdowns is

$$C = 3Y + Y^2$$

Show that $E[C] = 4\lambda + \lambda^2$.

Note that

$$\begin{aligned} E[C] &= E[3Y + Y^2] = 3E[Y] + E[Y^2] \\ &= 3\lambda + V(Y) + (E[Y])^2 \\ &= 3\lambda + \lambda + \lambda^2 \\ &= 4\lambda + \lambda^2 \end{aligned}$$

(c) Find a function of Y_1, Y_2, \dots, Y_n that is an unbiased estimator of C .

Note that by our computation in Part (A),

$$E[3Y_i + Y_i^2] = 4\lambda + \lambda^2 = E[C] \text{ for each } i=1, 2, \dots, n$$

Hence,

$$\frac{1}{n} \sum_{i=1}^n E[3Y_i + Y_i^2] = E[C]$$

$$\Rightarrow E\left[\frac{1}{n} \sum_{i=1}^n (3Y_i + Y_i^2)\right] = E[C]$$

Hence $\hat{E}[C] = \frac{1}{n} \sum_{i=1}^n (3Y_i + Y_i^2)$ is an

unbiased estimate of $E[C]$.