LECTURE - 11

CHEPTER-4

4.2 Variance and Covariance of random variables

The most important measure of variability of a random variable X is the variance of the random variable X or the variance of the probability distribution of X and is denoted by Var(X) or the symbol σ_X^2 , or simply by σ^2

Definition 4.3

Let X be a random variable with probability distribution f(x) and mean μ . The variance of X is

$$\sigma^2 = E[(X - \mu)^2] = \sum_x (x - \mu)^2 f(x), \text{ if X is discrete, and}$$

$$\sigma^2 = E[(X - \mu)^2] = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx, \text{ if X is continuous.}$$

The positive square root of the variance σ is called the standard deviation of X.

Theorem-4.2:

The variance of a random variable X is $\sigma^2 = E(X^2) - \mu^2$.

Theorem-4.3:

Let X be a random variable with probability distribution f(x). The variance of the random variable g(X) is

$$\sigma_{g(X)}^2 = E[g(X) - \mu_{g(X)}]^2 = \sum_x \left[g(x) - \mu_{g(X)}\right]^2 f(x)$$
 if X is discrete, and

$$\sigma_{g(X)}^2 = E\{[g(X) - \mu_{g(X)}]^2\} = \int_{-\infty}^{\infty} [g(x) - \mu_{g(X)}]^2 f(x) dx$$
 if X is continuous.

Definition 4.4

Let X and Y be random variables with joint probability distribution f(x, y). The covariance of X and Y is

$$\sigma_{XY} = E[(X - \mu_X)(Y - \mu_Y)] = \sum_x \sum_y (x - \mu_x)(y - \mu_y) f(x,y)$$
 if X and Y are discrete, and

$$\sigma_{XY} = E[(X - \mu_X)(Y - \mu_Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x - \mu_X)(y - \mu_Y)f(x, y)dxdy$$
 if X and Y are continuous

Theorem-4.4:

The covariance of two random variables X and Y with means μ_X and μ_Y , respectively, is given by

$$\sigma_{XY} = E(XY) - \mu_X \mu_Y$$
.

Example-4.9

Let the random variable X represents the number of defective parts for a machine when 3 parts are sampled from a production line and tested. The following is the probability distribution of X.

Calculate σ^2 .

Solution:

$$\mu = (0)(0.51) + (1)(0.38) + (2)(0.10) + (3)(0.01) = 0.61.$$

$$E(X^2) = (0)(0.51) + (1)(0.38) + (4)(0.10) + (9)(0.01) = 0.87.$$

Therefore,
 $\sigma^2 = 0.87 - (0.61)^2 = 0.4979.$

Example-4.10

The weekly demand for a drinking-water product, in thousands of liters, from a local chain of efficiency stores is a continuous random variable X having the probability density

$$f(x) = \begin{cases} 2(x-1) & 1 < x < 2\\ 0 & elsewhere \end{cases}$$

Find the mean and variance of X.

Solution:

Calculating E(X) and
$$E(X^2)$$
, we have $\mu = E(X) = 2 \int_1^2 x(x-1) dx = \frac{5}{3}$ and $E(X^2) = 2 \int_1^2 x^2(x-1) dx = \frac{17}{6}$ Therefore, $\sigma^2 = \frac{17}{6} - (\frac{5}{3})^2 = \frac{1}{18}$.

Exercise-4.34

Let X be a random variable with the following probability distribution:

$$\begin{array}{c|ccccc} x & -2 & 3 & 5 \\ \hline f(x) & 0.3 & 0.2 & 0.5 \\ \end{array}$$

Find the standard deviation of X.

Solution:

Calculating E(X) and $E(X^2)$, we have $\mu = (-2)(0.3) + (3)(0.2) + (5)(0.5) = 2.5$ $E(X^2) = (4)(0.3) + (9)(0.2) + (25)(0.5) = 15.5$ Therefore, $\sigma^2 = 15.5 - 2.5^2 = 9.25.$ The standard deviation $\sigma = \sqrt{9.25} = 3.04138$.

Exercise-4.35

The random variable X, representing the number of errors per 100 lines of software code, has the following probability distribution:

Solution:

E(X)=
$$\sum_{x} x f(x) = 2(0.01) + 3(0.25) + 4(0.4) + 5(0.3) + 6(0.04) = 4.11$$

$$E(X^{2}) = \sum_{x} x^{2} f(x) = 4(0.01) + 9(0.25) + 16(0.4) + 25(0.3) + 36(0.04) = 17.63$$

$$\sigma^{2} = E(X^{2}) - (E(X))^{2} = 17.63 - (4.11)^{2} = 0.7379$$

Exercise-4.50

For a laboratory assignment, if the equipment is working, the density function of the observed outcome X is

$$f(x) = \begin{cases} 2(1-x) & 0 < x < 1\\ 0 & elsewhere \end{cases}$$

Find the variance and standard deviation of X.

Solution:

Calculating E(X) and
$$E(X^2)$$
, we have $\mu = E(X) = 2 \int_0^1 x(1-x) dx = \frac{1}{3}$ and $E(X^2) = 2 \int_0^1 x^2 (1-x) dx = \frac{1}{6}$ Therefore, $\sigma^2 = \frac{1}{6} - (\frac{1}{3})^2 = \frac{1}{18}$ and $\sigma = \sqrt{\frac{1}{18}} = 0.2357$

Completed