Random Variable and Distribution Functions

Lesson 3: Moments and Expectation

Discrete Random Variable:

Let X is a discrete random variable which takes the values $x_1, x_2, x_3, \ldots, x_n$ with probabilities $p_1, p_2, p_3, \ldots, p_n$ such that the probability mass function is given by p(x), then

(i) Arithmetic mean:

$$\bar{x} = \sum_{i=1}^{n} x_i p_i$$
 or, $\sum x p(x)$

(ii) Moments:

- (a) r-th moment about any point x = A: $\mu_r' = \sum_r (x A)^r p(x)$
- (b) r-th moment about the origin: $\mu_r' = \sum_r x^r p(x)$
- (c) r-th moment about the mean: $\mu_r = \sum_r (x \bar{x})^r p(x)$

Continuous Random Variable:

Let f(x) be the probability density function of a random variable X defined between a and b, then

(i) Arithmetic mean:

$$\bar{x} = \int_{a}^{b} x f(x) dx$$

(ii) Moments:

- (a) r-th moment about any point x = A: $\mu_r' = \int_a^b (x A)^r f(x) dx$
- (b) r-th moment about the origin: $\mu_r' = \int_a^b x^r f(x) dx$

(c) r-th moment about the mean: $\mu_r = \int_a^b (x - \bar{x})^r f(x) dx$

(iii) Median:

If *M* is the median, then $\int_a^M f(x)dx = \int_M^b f(x)dx = \frac{1}{2}$. Solution of these two equations gives the median.

(iv) Mode:

Mode is the value of x for which f(x) is maximum. It is thus the solution of f'(x) = 0 with f''(x) < 0, provided it lies in [a, b]

Note: All other moment measures for skewness and kurtosis remain same

Problems:

Ex.1. A random variable X is defined as follows: P(X = 1) = p, P(X = 0) = 1 - p, 0 . Find mean and variance of the distribution.

Solution: We have the distribution given as

X	0	1
p(x)	1 <i>-p</i>	p

Here *X* is a discrete random variable, so the mean and the variance of the distribution are given by

$$\bar{x} = \sum_{i=1}^{2} x_i p_i = 0 \times (1-p) + 1 \times p = p$$

$$\mu_2 = {\mu_2}' - {\mu_1}'^2 = \sum_{i=1}^2 x_i^2 p_i - p^2 = p - p^2 = p(1-p)$$

Ans.

Ex.2. Find mean and standard deviation of the distribution with density function $f(x) = kx^2e^{-x}$, $0 < x < \infty$

Solution: For the density function f(x), we always have

$$\int_{-\infty}^{\infty} f(x)dx = 1 \to k \int_{0}^{\infty} x^{2}e^{-x}dx = 1 \to k\Gamma(3) = 1 \to k = \frac{1}{\Gamma(3)} = \frac{1}{2\Gamma(1)} = \frac{1}{2}$$

Therefore mean of the distribution is given by

$$\bar{x} = \int_{0}^{\infty} x \left(\frac{x^2}{2} e^{-x} \right) dx = \frac{1}{2} \int_{0}^{\infty} x^3 e^{-x} dx = \frac{1}{2} \Gamma(4) = 3$$

Also the variance of the distribution will be

$$\mu_2 = {\mu_2}' - {\mu_1}'^2 = \int_0^\infty x^2 \left(\frac{x^2}{2}e^{-x}\right) dx - 3^2 = 12 - 9 = 3$$

Hence the standard deviation of the distribution will be $\sqrt{3}$.

Ans.

Ex.3. The probability density function of a random variable *X* is $f(x) = \frac{\pi}{10} \sin\left(\frac{\pi x}{5}\right)$, $0 \le x \le 5$. obtain the median of the distribution.

Solution: Let us consider that M is the median of the distribution. Then we have

$$\int_{0}^{M} \frac{\pi}{10} \sin\left(\frac{\pi x}{5}\right) dx = \frac{1}{2} \to \int_{0}^{M} \sin\left(\frac{\pi x}{5}\right) dx = \frac{5}{\pi} \to M = 2.5$$

Ans.

Mathematical Expectation:

The average value of a random phenomenon is termed as its mathematical expectation or expected values.

(i) For a discrete random variable X with mass function p(x)

$$E(X) = \sum_{x} x p(x)$$

provided the series is absolutely convergent, i.e, $\sum_{x} |x| p(x) < \infty$

(ii) For a continuous random variable X with density function f(x)

$$E(X) = \int_{-\infty}^{\infty} x f(x) dx$$

provided the integral is absolutely convergent, i.e, $\int_{-\infty}^{\infty} |x| f(x) dx < \infty$

Expected value of a function of a random variable:

Let X is a random variable with mass function p(x) / density function f(x). Let g(.) is a function such that g(X) is a random variable. Then expected value of g(X) is defined as

$$E[g(X)] = \sum_{x} g(x)p(x) - - - \text{(discrete)}$$
$$= \int_{-\infty}^{\infty} g(x)f(x)dx - - \text{(continuous)}$$

Particular case:

1. If
$$g(X) = X^r$$
, then $E[X^r] = \sum_x x^r p(x)$ ----- (discrete)
$$= \int_{-\infty}^{\infty} x^r f(x) dx ---- \text{ (continuous)}$$

$$= \mu_r' = \text{r-th moment about the origin}$$

The mean and variance of the distribution can be found out from this expression.

2. If
$$g(X) = [X - E(X)]^r$$
, then
$$E([X - E(X)]^r) = \sum_x (x - \bar{x})^r p(x) ---- \text{(discrete)}$$

$$= \int_{-\infty}^{\infty} (x - \bar{x})^r f(x) dx ---- \text{(continuous)}$$

$$= \mu_r = \text{r-th moment about the mean}$$

3. If
$$g(X) = c$$
, a constant, then $E(c) = \int_{-\infty}^{\infty} cf(x) dx = c$

Properties of mathematical expectation:

1. Addition Theorem:

For two random variables X and Y and two constants a and b, we have E(aX + bY) = aE(X) + bE(Y)

2. <u>Multiplication Theorem</u>:

If X and Y are independent random variables, then E(XY) = E(X)E(Y)

- 3. If X and g(X) are both random variables and a is a constant, then
- (i) E[ag(X)] = aE[g(X)] (ii) E[g(X) + a] = E[g(X)] + a
- 4. If X is a random variable ≥ 0 , then $E(X) \geq 0$
- 5. If X and Y are two random variables and X < Y, then $E(X) \le E(Y)$
- 6. For a random variable X, $|E(X)| \le E(|X|)$

7. If *X* and *Y* are two independent random variables, then

$$E[h(X)k(Y)] = E[h(X)]E[k(Y)]$$

- 8. If X is a random variable, then $Var(aX + b) = a^2Var(X)$, where a and b are constants
- 9. If X and Y are two random variables, then covariance between them is given by

$$Cov(X,Y) = E(XY) - E(X)E(Y)$$

Problems:

Ex.1. The probability distribution of a commodity is given below:

Demand (X_i)	5	6	7	8	9	10
Probability (p_i)	0.05	0.1	0.3	0.4	0.1	0.05

Find the expected demand and its variance.

Solution: A discrete distribution is given, hence

$$E(X) = \sum_{x} xp(x) = 7.55$$

And

$$Var(X) = E(X^2) - \{E(X)\}^2 = \sum_{x} x^2 p(x) - 7.55^2 = 1.25$$

Ans.