

WEEK 12

Expectations For Random Variables

Definition:

The **expected value** of a random variable is the mean value of the variable X in the sample space, or population, of possible outcomes. *Expected value*, denoted by $E(X)$, can also be interpreted as the mean value that would be obtained from an infinite number of observations on the random variable.

$$E(X) = \begin{cases} \sum_{i=1}^n x_i P(X = x_i) & \text{for discrete random variable} \\ \int_{-\infty}^{\infty} x f(x) dx & \text{for continuous random variable} \end{cases}$$

Expected Values of functions of r.v.

If $h(X)$ is a function of X , then

$$\mu_{h(X)} = E[h(X)] = \int_{-\infty}^{\infty} h(x) \cdot f(x) dx$$

$$E[h(X)] = E[aX + b] = aE[X] + b$$

Properties of Expectation

Property 1. Addition Theorem of Expectation

If X and Y are random variables then $E(X + Y) = E(X) + E(Y)$, provided all the expectation exists.

Property 2: Multiplication theorem of Expectation

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

Property 3: If X is a random variable and 'a' is constant.

- (i) $E[a H(X)] = a E[H(X)]$
- (ii) $E[H(X) + a] = E[H(X)] + a$

Where H(X) is a function of X, is a r.v and all the expectation are exists.

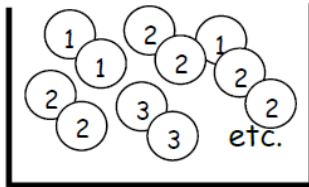
4.2 Variance

The variance of a random variable X is defines as

$$Var(X) = E(X^2) - (E(X))^2$$

Example:

Consider a population consisting of 100 families in a community. Suppose that 30 families have just 1 child, 50 families have 2 children, and 20 families have 3 children. What is the mean or average number of children per family for this population?



Population of 100 families

Mean = (sum of all values)/100

$$= [1(30) + 2(50) + 3(20)]/100$$

$$= 1(30/100) + 2(50/100) + 3(20/100)$$

$$= 1(0.30) + 2(0.50) + 3(0.20)$$

$$= 1.9 \text{ children per family}$$

Mean = Sum of (value x probability of that value)

Example: 4.1 Find the expectation and variance of the number on a die when thrown

Solution

Let X be a random variable representing the number on a die when thrown. Then X can take any one of the values 1,2,3,4,5,6 each with equal probability $1/6$

x	1	2	3	4	5	6
$P(X=x)$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$

$$\begin{aligned} E(X) &= \sum_{i=1}^6 x_i P(X = x_i) = 1\frac{1}{6} + 2\frac{1}{6} + 3\frac{1}{6} + 4\frac{1}{6} + 5\frac{1}{6} + 6\frac{1}{6} \\ &= \frac{1+2+3+4+5+6}{6} \\ E(X) &= \frac{21}{6} \end{aligned}$$

Example 4.2 If a pair of fair dice is tossed and X denotes the sum of the numbers on them, find the expectation of X.

Solution: Clearly X may be at least 2 and at most 12

X	2	3	4	5	6	7	8	9	10	11	12
P(X)	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$

$$\begin{aligned}
 E(X) &= \sum_{i=2}^{12} x_i P(X = x_i) = 2 \frac{1}{36} + 3 \frac{2}{36} + 4 \frac{3}{36} + 5 \frac{4}{36} + 6 \frac{5}{36} + 7 \frac{6}{36} + 8 \frac{5}{36} \\
 &\quad + 9 \frac{4}{36} + 10 \frac{3}{36} + 11 \frac{2}{36} + 12 \frac{1}{36} \\
 &= \frac{1}{36} [2 + 6 + 12 + 20 + 30 + 42 + 48 + 36 + 30 + 22 + 12]
 \end{aligned}$$

$$E(X) = \frac{252}{36} = 7$$

Example 4.3 If X be a random variable with the following probability distribution

X	-3	6	9
P(x)	$\frac{1}{6}$	$\frac{1}{2}$	$\frac{1}{3}$

Find $E(X)$, $E(X^2)$ and $E(2X+1)^2$

Solution

$$E(X) = \sum x_i P(X = x_i) = -3 \frac{1}{6} + 6 \frac{1}{2} + 9 \frac{1}{3} = \frac{-3 + 18 + 18}{6} = \frac{33}{6} = \frac{11}{2}$$

$$E(X) = \frac{11}{2}$$

$$E(X^2) = \sum x_i^2 P(X = x_i) = (-3)^2 \frac{1}{6} + 6^2 \frac{1}{2} + 9^2 \frac{1}{3} = \frac{93}{2}$$

$$E(X^2) = \frac{93}{2}$$

$$E(2X+1)^2 = E[4X^2 + 4X + 1] = E[4X^2] + E[4X] + E[1]$$

$$= 4E[X^2] + 4E[X] + 1$$

$$= 4 \cdot \frac{93}{2} + 4 \cdot \frac{11}{2} + 1 = 209$$

$$E(2X+1)^2 = 209$$

Example: 4.4 In a continuous distribution the probability density function of X is

$$f(x) = \begin{cases} \frac{3}{4}x(2-x) & , 0 < x < 2 \\ 0 & , \text{otherwise} \end{cases} \quad \text{Find the expectation of the distribution.}$$

Solution.

$$E(X) = \int_0^2 x f(x) dx = \int_0^2 x \cdot \frac{3}{4}x(2-x) dx$$

$$= \frac{3}{4} \int_0^2 x^2(2-x) dx = \frac{3}{4} \int_0^2 2x^2 - x^3 dx$$

$$= \frac{3}{4} \left[2 \frac{x^3}{3} - \frac{x^4}{4} \right]_0^2 = \frac{3}{4} \left[\left(2 \frac{2^3}{3} - \frac{2^4}{4} \right) - \left(2 \frac{0^3}{3} - \frac{0^4}{4} \right) \right]$$

$$= \frac{3}{4} \left[2 \frac{8}{3} - \frac{16}{4} \right] = \frac{3}{4} \left[\frac{16}{3} - \frac{16}{4} \right]$$

$$= \frac{3}{4} \left[\frac{16}{3} - 4 \right] = \frac{3}{4} \left[\frac{16-12}{3} \right] = \frac{3}{4} \left[\frac{4}{3} \right] = 1$$

$$E(X) = 1$$

Example 7.10 (a) What is the mathematical expectation of the number of heads when 3 coins are tossed?

(b) What is the expectation of the number of failures preceding the first success in an infinite sequence of independent trials with constant probability of success? (P.H., D.M., 1980)

a) Let the r.v. X represent the number of heads when three fair coins are tossed. Then X has the following p.d.

x	0	1	2	3
$f(x)$	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$

Hence the mathematical expectation of X is

$$E(X) = 0 \times \frac{1}{8} + 1 \times \frac{3}{8} + 2 \times \frac{3}{8} + 3 \times \frac{1}{8} = \frac{12}{8} = 1.5$$

It should be noted that $E(X)$ is 1.5, which is not an integer and not any value X could actually have. This means that, if the experiment of tossing 3 coins is repeated a very large number of times, we expect on the average to get 1.5 heads.

- 5) Let the r.v. X denote the number of failures preceding the first success. Then as X takes the value 0, 1, 2, 3, ..., the respective probabilities are p, qp, q^2p, q^3p, \dots where $q = 1 - p$.

$$\begin{aligned} \text{Hence } E(X) &= x_1 f(x_1) + x_2 f(x_2) + x_3 f(x_3) + \dots \\ &= 0 \cdot p + 1 \cdot qp + 2 \cdot q^2p + 3 \cdot q^3p + \dots \\ &= qp(1 + 2q + 3q^2 + \dots) \\ &= qp(1 - q)^{-2} = qp(p)^{-2} = \frac{q}{p} \end{aligned}$$

Example 7.12 Find the expected value of the r.v. X having the p.d.f.

$$\begin{aligned} f(x) &= 2(1 - x), \quad 0 < x < 1 \\ &= 0, \quad \text{elsewhere,} \end{aligned}$$

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

$$= 2 \int_0^1 x(1 - x) dx$$

$$= 2 \left[\frac{x^2}{2} - \frac{x^3}{3} \right]_0^1 = 2 \left[\frac{1}{2} - \frac{1}{3} \right] = \frac{1}{3}$$

Definitions:

Mean and standard deviation of a discrete random variable

Suppose that X is a discrete random variable with possible values x_1, x_2, x_3, \dots occurring with probabilities p_1, p_2, p_3, \dots , then

the expected value (or mean) of X is given by $\mu = E(X) = \sum x_i p_i$

the *variance* of X is given by $V(X) = \sigma^2 = \sum (x_i - \mu)^2 p_i$

and so the *standard deviation* of X is given by $\sigma = \sqrt{\sum (x_i - \mu)^2 p_i}$

The sums are taken over all possible values of the random variable X .

Variance

The **variance** of a continuous random variable X with pdf $f(x)$ and mean value μ is

$$\begin{aligned}\sigma_X^2 = V(X) &= \int_{-\infty}^{\infty} (x - \mu)^2 \cdot f(x) dx \\ &= E[(X - E(X))^2] \\ &= E(X^2) - E(X)^2\end{aligned}$$

The **standard deviation** (SD) of X is

$$\sigma_X = \sqrt{V(X)}$$

When $h(X) = aX + b$,

$$V[h(X)] = V[aX + b] = a^2 \cdot \sigma_X^2 \text{ and } \sigma_{aX+b} = |a| \cdot \sigma_X$$

We call this expected value the variance and denote it by $\text{Var}(X)$ or σ^2 . That is $\sigma^2 = E(X - \mu)^2 = E(X^2) - [E(X)]^2$. The positive square root of the variance, as before, is called the standard deviation.

It is useful to note the following important results about variance.

- i) $\text{Var}(X)$ cannot be negative.
- ii) $\text{Var}(a) = 0$, where a is a constant.
- iii) $\text{Var}(aX) = a^2 \text{Var}(X)$, where a is a constant.
- iv) $\text{Var}(aX + b) = a^2 \text{Var}(X)$, where a and b are constants.

Example 7.13 Let X have the following probability distribution:

x_i	1	2	3	4	5
$f(x_i)$	0.2	0.3	0.2	0.2	0.1

Find the probability functions of $3X - 1$, X^2 and $X^2 + 2$; and find $E(3X - 1)$, $E(X^2)$ and $E(X^2 + 2)$.

The probability distribution of the r.v. $H(X) = 3X - 1$, is

Values of X ,	x_i	1	2	3	4	5
Probabilities,	$f(x_i)$	0.2	0.3	0.2	0.2	0.1
Values of $3X - 1$, $(3x_i - 1)$		2	5	8	11	14

$$E(3X - 1) = \sum H(x_i) f(x_i) = \sum (3x_i - 1) f(x_i)$$

$$= 2 \times 0.2 + 5 \times 0.3 + 8 \times 0.2 + 11 \times 0.2 + 14 \times 0.1$$

$$= 0.4 + 1.5 + 1.6 + 2.2 + 1.4 = 7.1$$

The p.d. of $H(X) = X^2$ is

x_i	1	2	3	4	5
$f(x_i)$	0.2	0.3	0.2	0.2	0.1
x_i^2	1	4	9	16	25

and

$$\begin{aligned}
 E(X^2) &= \sum x_i^2 f(x_i) \\
 &= 1 \times 0.2 + 4 \times 0.3 + 9 \times 0.2 + 16 \times 0.2 + 25 \times 0.1 \\
 &= 0.2 + 1.2 + 1.8 + 3.2 + 2.5 = 8.9
 \end{aligned}$$

Similarly,

$$\begin{aligned}
 E(X^2 + 2) &= \sum (x_i^2 + 2) f(x_i) \\
 &= 3 \times 0.2 + 6 \times 0.3 + 11 \times 0.2 + 18 \times 0.2 + 27 \times 0.1 \\
 &= 0.6 + 1.8 + 2.2 + 3.6 + 2.7 = 10.9
 \end{aligned}$$

Example 7.14 Let X be a r.v. with p.d.f.

$$\begin{aligned}
 f(x) &= 2(x-1), & 1 < x < 2 \\
 &= 0, & \text{elsewhere.}
 \end{aligned}$$

Find the expected values of $H(X) = 2X - 1$ and $H(X) = X^2$.

Now

$$\begin{aligned}
 E(2X - 1) &= \int_{-\infty}^{\infty} (2x - 1) f(x) dx = 2 \int_1^2 (2x - 1)(x - 1) dx \\
 &= 2 \int_1^2 (2x^2 - 3x + 1) dx = 2 \left[\frac{2x^3}{3} - \frac{3x^2}{2} + x \right]_1^2 \\
 &= 2 \left[\left(\frac{16}{3} - 6 + 2 \right) - \left(\frac{2}{3} - \frac{3}{2} + 1 \right) \right] \\
 &= 2 \left[\frac{4}{3} - \frac{1}{6} \right] = \frac{7}{3}
 \end{aligned}$$

$$\text{and } E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) dx$$

$$= 2 \int_1^2 x^2 (x-1) dx = 2 \left[\frac{x^4}{4} - \frac{x^3}{3} \right]_1^2$$

$$= 2 \left[\left(4 - \frac{8}{3} \right) - \left(\frac{1}{4} - \frac{1}{3} \right) \right]$$

$$= 2 \left[\frac{4}{3} + \frac{1}{12} \right] = \frac{17}{6}$$

Example 7.15 If the continuous r.v. X has p.d.f.

$$f(x) = \frac{3}{4} (3-x)(x-5), \quad 3 \leq x \leq 5$$

$$= 0, \quad \text{elsewhere.}$$

Find the arithmetic mean, variance and standard deviation of X .

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

$$= \frac{3}{4} \int_3^5 x(3-x)(x-5) dx$$

$$= \frac{3}{4} \int_3^5 (-x^3 + 8x^2 - 15x) dx = \frac{3}{4} \left[-\frac{x^4}{4} + \frac{8x^3}{3} - \frac{15x^2}{2} \right]_3^5$$

$$= \frac{3}{4} \left[\left(-\frac{625}{4} + \frac{1000}{3} - \frac{375}{2} \right) - \left(-\frac{81}{4} + \frac{216}{3} - \frac{35}{2} \right) \right]$$

$$= \frac{3}{4} \left[\left(-\frac{125}{12} + \frac{63}{4} \right) \right] = \frac{3}{4} \left(\frac{64}{12} \right) = 4$$

Again $E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) dx$

$$= \frac{3}{4} \int_3^5 x^2 (3-x)(x-5) dx$$

$$= \frac{3}{4} \int_3^5 (-x^4 + 8x^3 - 15x^2) dx = \frac{3}{4} \left[-\frac{x^5}{5} + \frac{8x^4}{4} - \frac{15x^3}{3} \right]_3^5$$

$$= \frac{3}{4} \left[\frac{-1}{5} (3125) + 2(625) - 5(125) - \left\{ \frac{-1}{5} (243) + 2(81) - 5(27) \right\} \right]$$

$$= \frac{3}{4} \left[0 + \frac{243}{5} - 162 + 135 \right] = \frac{3}{4} \left[\frac{108}{5} \right] = \frac{81}{5}$$

$$\therefore \text{Var}(X) = E(X^2) - [E(X)]^2$$

$$= \frac{81}{5} - (4)^2 = \frac{1}{5} = 0.2, \text{ and}$$

$$S.D.(X) = \sqrt{0.2} = 0.447$$

∴ the mean = 4, variance = 0.2 and standard deviation = 0.447.

Example 7.16 The continuous r.v. X has p.d.f. $f(x)$, where $f(x) = \frac{3}{4}(1+x^2)$ for $0 \leq x \leq 1$. $E(X) = \mu$ and $Var(X) = \sigma^2$, find $P(|X - \mu| < \sigma)$.

$$\text{Now } E(X) = \int_{-\infty}^{\infty} xf(x) dx$$

$$= \frac{3}{4} \int_0^1 x(1+x^2) dx = \frac{3}{4} \int_0^1 (x+x^3) dx$$

$$= \frac{3}{4} \left[\frac{x^2}{2} + \frac{x^4}{4} \right]_0^1 = \frac{3}{4} \left[\frac{1}{2} + \frac{1}{4} \right] = \frac{9}{16} = 0.5625$$

$$\text{And } E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) dx$$

$$= \frac{3}{4} \int_0^1 x^2(1+x^2) dx = \frac{3}{4} \int_0^1 (x^2 + x^4) dx$$

$$= \frac{3}{4} \left[\frac{x^3}{3} + \frac{x^5}{5} \right]_0^1 = \frac{3}{4} \left[\frac{1}{3} + \frac{1}{5} \right]$$

$$= \frac{3}{4} \left(\frac{8}{15} \right) = \frac{2}{5} = 0.4$$

$$\therefore Var(X) = E(X^2) - [E(X)]^2 = \frac{2}{5} - \left(\frac{9}{16} \right)^2$$

$$= \frac{107}{1280} = 0.0836, \text{ so that}$$

$$\text{S.D. (X) or } \sigma = \sqrt{0.0836} = 0.289$$

$$\text{Now } P(|X - \mu| < \sigma) = P(-\sigma < X - \mu < \sigma)$$

$$= P(\mu - \sigma < X < \mu + \sigma)$$

$$= P(0.5625 - 0.289 < X < 0.5625 + 0.289)$$

$$= P(0.2735 < X < 0.8515), \text{ and}$$

$$P(0.2735 < X < 0.8515) = \frac{3}{4} \int_{0.2735}^{0.8515} (1 + x^2) dx$$

$$= \frac{3}{4} \left[x + \frac{x^3}{3} \right]_{0.2735}^{0.8515}$$

$$= \frac{3}{4} \left[0.8515 + \frac{(0.8515)^3}{3} - \left(0.2735 + \frac{(0.2735)^3}{3} \right) \right]$$

$$= \frac{3}{4} [1.05729 - 0.28032] = 0.8527$$

$$\therefore P(|X - \mu| < \sigma) = 0.8527.$$