(I)

Mathematical Expectation. Let X be a random variable (r.v.) with p.d.f. (p.m.f.) f(x). Then its mathematical expectation, denoted by E(X) is given by:

$$E(X) = \int_{x}^{\infty} x f(x) dx, \text{ (for continuous } r.v.) = \text{mean} = \overline{X}$$
$$= \sum_{x}^{\infty} x f(x), \text{ (for discrete } r.v.)$$

Remarks.

- E(X) exists iff E[X] exists.
- The expectation of a random variable is thought of as a long-term average.

Expectation of a Function of a Random Variable. Consider a r.v. X with p.d.f. (p.m.f.) f(x) and distribution function F(x). If $g(\cdot)$ is a function such that g(X) is a r.v. and E[g(X)] exists (i.e., is defined), then

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

 $= \sum_{x=0}^{\infty} g(x) f(x)$ (For continuous r.v.)

(For discrete r.v.)

$$E(X) = \sum \pi p(\pi) = \sum \pi F(\pi)$$

$$E(g(n)) = \sum g(n) f(n)$$

$$= \int g(x) f(x)$$

d is DRV

n in (RV

g(n)=(2n+5) $E(2n+5) = \sum_{n=0}^{\infty} (2n+5) p(n) = \sum_{n=0}^{\infty} (2n+5) p(n)$ = (2n+5) p(n)



Addition Theorem of Expectation

If X and Y are random variables then

$$E(X+Y)=E(X)+E(Y),$$

provided all the expectations exist.

$$E(X)=2$$

$$E(X+Y)=E(X)+E(Y)$$

$$= 2+3=5$$

The mathematical expectation of the sum of n random variables is equal to the sum of their expectations, provided all the expectations exist.

Symbolically, if $X_1, X_2, ..., X_n$ are random variables then

$$E(X_1 + X_2 + ... + X_n) = E(X_1) + E(X_2) + ... + E(X_n)$$

or $E\left(\begin{array}{c}\sum_{i=1}^{n}X_{i}\\\end{array}\right)=\sum_{i=1}^{n}E\left(\dot{X}_{i}\right),$

if all the expectations exist.

$$E(x_1+y_2+x_3) \cdots x_n) = E(x_1) + E(x_2) + \cdots + E(x_n)$$

$$E\left(\sum_{i=1}^n x_i\right) = \sum_{i=1}^\infty E(x_1)$$



Multiplication Theorem of Expectation

If X and Y are independent random variables, then

$$E(XY) = E(X) \cdot E(Y)$$

$$E(x) = 2$$

$$E(xy) = E(x) E(y)$$

$$= 2 \times 3$$

$$= 6$$

VI

Generalisation to n-variables.

The mathematical expectation of the product of a number of independent random variables is equal to the product of their expectations. Symbolically, if $X_1, X_2, ..., X_n$ are n independent random variables, then

$$E(X_1 X_2 ... X_n) = E(X_1) E(X_2) ... E(X_n)$$

i.e.,
$$E\left(\begin{array}{cc} \prod_{i=1}^{n} X_{i} \end{array}\right) = \prod_{i=1}^{n} E\left(X_{i}\right)$$

provided all the expectations exist.

$$E(x_1 x_2 \dots x_n) = E(x_1) E(x_2) \dots E(x_n)$$

$$E(\prod_{i=1}^n x_i) = \prod_{i=1}^n E(x_i)$$

W

If X is a random variable and 'a' is constant, then

(i)
$$E[a\Psi(X)] = a E[\Psi(X)]$$

(ii)
$$E[\Psi(X) + a] = E[\Psi(X)] + a$$
,

where $\Psi(X)$, a function of X, is a r.v. and all the expectations exist.

$$E(aY(n)) = aE(Y)$$

$$E\left(\psi(m)+\alpha\right)=$$

$$= E(\Psi(m)) + E(a)$$

$$= E(\Psi x) + \alpha$$

$$E(a) = \sum a p(a) = a \sum p(a)$$

$$E(\alpha) = \int_{-\alpha}^{\alpha} u \, \rho(x) \, dx = \alpha \int_{-\alpha}^{\alpha} \rho(x) \, dx$$

$$= a(1)$$

If X is a random variable and a and b are constants, then

$$E(aX + b) = a E(X) + b$$

provided all the expectations exist.

$$E(2n+5) = E(2x) + E(5)$$

= 2 $E(x) + 5$

Expectation of a Linear Combination of Random Variables

Let $X_1, X_2, ..., X_n$ be any n random variables and if $a_1, a_2, ..., a_n$ are any n constants, then

$$E\left(\sum_{i=1}^{n} a_{i} X_{i}\right) = \sum_{i=1}^{n} a_{i} E\left(X_{i}\right)$$

provided all the expectations exist. /

$$F(a_1 \times_1 + a_2 \times_2 + a_3 \times_3 + \cdots + a_n \times_n)$$

$$= \alpha E(x_1) + \alpha_2 E(x_2) + \alpha_3 E(x_3)$$

If
$$X \ge 0$$
 then $E(X) \ge 0$.

$$\sum x b(x) > 0$$

Let X and Y be two random variables such that
$$Y \le X$$
 then

$$E(Y) \leq E(X)$$
provided the expectations exist.

$$|E(X)| \le E|X|$$
 provided the expectations exist.

al

Let X be a random variable with the following probability

distribution:

$$x$$
 : -3 6 9
 $P_r(X = x)$: $1/6$ 1/2 1/3

Find E(X) and $E(X^2)$ and using the laws of expectation, evaluate $E(2X+1)^2$.

$$E(2x+1)^{2} = E[(2x)^{2}+1^{2}+2(2x)]$$

$$= E[4x^{2}+1+4x]$$

$$= 4E(x^{2})+E(1)+4E(x)$$

$$= 24(93)+1+22$$

$$= 186+1+22$$

$$= 269$$

$$V(aX+b) = a^2V(X)$$

where a and b are constants.

(i) If
$$b = 0$$
, then $V(a X) = a^2 V(X)$

> Variance is not independent of change of scale.

(ii) If
$$a = 0$$
, then $V(b) = 0$

⇒ Variance of a constant in zero.

(iii) If
$$a = 1$$
, then $V(X + b) = V(X)$

⇒ Variance is independent of change of origin.

$$A_{(X)} = E(x_3) - (E(x))_5$$

$$A_{(X)} = E(x_3) - (E(x))_5$$

$$x = 2in + 5$$

$$V(ux) + V(s)$$

$$= 4^{2} V(x) + 0$$

$$= 16 V(x)$$

Given the following table:

.[х	-3	-2	'-1,	0	1	2	, 3
	p (x)	0.05	0¦10	0.30	0	0·30	0.15	0.10

Compute

(i) E(X),

(ii) $E(2X \pm 3)$, (iii) E(4X + 5),

(iv) $E(X^2)$

(v) V(X), and (vi) $V(2X \pm 3)$.

$$E(2x+3) = E(2x) + E(3)$$

$$= 2E(X) + 3$$

$$= 2(0.15) + 3 = 0.5 + 3 = 3.5$$

$$E(2X-3) = E(3)$$

$$= 2F(x) - 3$$

$$=2(0.25)-3=0.5-3=-2.5$$

$$E(ux+s) = ME(x) + S$$

$$= u(0.2s) + S$$

$$= 1+S = G$$

E(X)=0.52

$$V(x) = E(x^{2}) - E(x)^{2}$$

$$= 2.95 - (0.25)^{2}$$

$$= 2.8875$$

$$V(2x+3) = 2^2 V(x) = 4(2.8875) = 11.57$$

 $V(2x-3) = 2^2 V(x) = 4(2.8875) = 11.55$