Mathematical Expectation or Expected Value of a random variable

Let X be a rv. Then the mathematical expectation of X is defined as

$$E(X) = \begin{cases} \sum_{i=1}^{\infty} x_i p(x_i) \text{ ; if x is a drv.} \\ \int_{-\infty}^{\infty} x f(x) dx \text{ ; if x is a crv} \end{cases}$$

Properties:

(i).
$$E(C) = C$$

(ii).
$$E(CX) = CE(X)$$

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

Generalization:

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

i.e. $E(\sum_{i=1}^n X_i) = \sum_{i=1}^n E(X_i)$

(iv).
$$E(XY) = E(X).E(Y)$$
 iff X and Y are independent

Variance:

Let X be a rv. Then the variance of X, denoted by σ^2 , is given by

$$V(X) = E(X - E(X))^{2},$$

= E(X²) - (E(X))²

Properties:

(i).
$$V(C)=0$$

(ii).
$$V(CX) = C^2V(X)$$

(iii).
$$V(X + Y) = V(X) + V(Y)$$
 iff X and Y independent

$$= E(X + Y)^2 - (E(X + Y))^2$$

$$= E(X^2) + E(Y^2) + 2E(XY) - (E(X))^2 - (E(Y))^2 - 2E(X)E(Y)$$

$$= E(X^2) - (E(X))^2 + E(Y^2) - (E(Y))^2 + 2[E(XY) - E(X)E(Y)]$$

$$= V(X) + V(Y) + 0$$

$$\therefore V(X+Y) = V(X) + V(Y)$$

Also $V(aX+bY) = a^2V(X)+b^2V(Y)$ iff X and Y are independent

Covariance of (XY)

$$Cov(XY) = E[(X-E(X))(Y-E(Y))]$$

$$= E(XY)-E(X).E(Y)$$

Note 1: If X and Y are independent then Cov(XY)=0. However, Cov(XY)=0 does not necessarily mean that the rvs are independent.

Note 2: If X and Y are any two rvs then we have

$$V(X+Y)=V(X)+V(Y)+2[E(XY)-E(X)E(Y)]=V(X)+V(Y)+2Cov(XY)$$

Also,
$$V(aX + bY) = a^2V(X) + b^2V(Y) + 2ab Cov(XY)$$

Mean and Variance of Binomial distribution:

Let
$$X \sim B(n,p)$$

$$P(X = x) = \binom{n}{x} p^{x} q^{n-x}; x=0,1,2,...,n$$

$$E(X) = \sum_{x=0}^{n} xp(x) = \dots = np$$

Thus mean of B.D is np

Now,
$$V(X) = E(X^2) - (E(X))^2$$

where,
$$E(X^2) = \sum_{x=0}^{n} x^2 \binom{n}{x} p^x q^{n-x} = \dots = n(n-1)p^2 + np$$

$$V(X) = n(n-1)p^{2} + np - (np)^{2}$$

$$= npq$$

Thus for the B.D. mean > variance; for 0

Mean and variance of Uniform distribution:

Let X~U[a,b]

We have
$$f(x) = \begin{cases} \frac{1}{b-a}; & a \le x \le b \\ 0; & \text{elsewhere} \end{cases}$$

The mean of uniform distribution is

$$E(X) = \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_{a}^{b} x \frac{1}{b-a} dx = \frac{a+b}{2}$$

Now,
$$V(X) = E(X^2) - (E(X))^2$$

where,
$$E(X^2) = \int_a^b x^2 \frac{1}{b-a} dx = \frac{1}{3} (b^2 + ab + a^2)$$

Thus
$$V(X) = \frac{1}{3}(b^2 + ab + a^2) - \frac{(a+b)^2}{4}$$

$$V(X) = \frac{1}{12}(b-a)^2$$