

Expectation of a Sum of Random variables:***Theorem:***

If X and Y are random variables, then

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

Proof:

$$\begin{aligned} E(X + Y) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x + y) f(x, y) dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f(x, y) dx dy + \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f(x, y) dx dy \\ &= \int_{-\infty}^{\infty} x \left(\int_{-\infty}^{\infty} f(x, y) dy \right) dx + \int_{-\infty}^{\infty} y \left(\int_{-\infty}^{\infty} f(x, y) dx \right) dy \\ &= \int_{-\infty}^{\infty} x f(x) dx + \int_{-\infty}^{\infty} y f(y) dy = E(X) + E(Y) \end{aligned}$$

A similar result can be shown in the discrete case.

In general, for any n ,

$$E(X_1 + X_2 + \dots + X_n) = E(X_1) + E(X_2) + \dots + E(X_n)$$

Expectation of a Product of Independent Random Variables:***Theorem:***

If X and Y are two independent random variables, then

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

Proof:

$$\begin{aligned}
E(XY) &= \sum_x \sum_y xyf(x, y) = \sum_x \sum_y xyf(x)f(y) \\
&= \sum_x xf(x) \sum_y yf(y) \\
&= E(X)E(Y)
\end{aligned}$$

Covariance and Correlation Coefficient:

Def: If X and Y are two random variables with respective means \bar{X} and \bar{Y} , then the covariance between X and Y is

$$\text{cov}(X, Y) = E(X - \bar{X})(Y - \bar{Y})$$

Note: If X and Y are independent random variables, then
 $\text{cov}(X, Y) = E(X - \bar{X})E(Y - \bar{Y}) = 0$

Since

$$E(X - \bar{X}) = E(X) - E(\bar{X}) = \bar{X} - \bar{X} = 0$$

A useful expression for $\text{cov}(X, Y)$ can be obtained by expanding the right side of the definition. This yields:

$$\begin{aligned}
\text{cov}(X, Y) &= E(X - \bar{X})(Y - \bar{Y}) = E(X - \mu_x)(Y - \mu_y) \\
&= E(XY - \mu_x Y - \mu_y X + \mu_x \mu_y) \\
&= E(XY) - \mu_x E(Y) - \mu_y E(X) + \mu_x \mu_y \\
&= E(XY) - 2\mu_x \mu_y + \mu_x \mu_y = E(XY) - E(X)E(Y)
\end{aligned}$$

From its definition, we see that:

$$\text{cov}(X, Y) = \text{cov}(Y, X) \text{ and}$$

$$\text{cov}(X, X) = \text{var}(X)$$

Def: The ratio

$$\rho_{XY} = \frac{\text{cov}(X, Y)}{\sqrt{\text{var}(X) \text{var}(Y)}} = \frac{E(X - \bar{X})(Y - \bar{Y})}{\sqrt{E(X - \bar{X})^2} \sqrt{E(Y - \bar{Y})^2}}$$

is defined as the correlation coefficient between X and Y .