Properties of Expectations

Expectation of Sums of Random Variables [Ross S7.2]

Recall that the mean value of X is

$$E[X] = \begin{cases} \sum_x x p_X(x) & X \text{ is discrete} \\ \\ \int_{-\infty}^{\infty} x f_X(x) dx & X \text{ is continuous} \end{cases}$$

Proposition 30.1 Let X and Y be two random variables. Let g(x,y) be a function. Then

$$E[g(X,Y)] = \begin{cases} \sum_{y} \sum_{x} g(x,y) p_{XY}(x,y) & X, Y \text{ are discrete} \\ \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f_{XY}(x,y) dx dy & X, Y \text{ are continuous} \end{cases}$$

Why?

[Only show for continuous case and g(x,y) is non-negative]

Recall from Proposition 16.2:

$$E[Z] = \int_0^\infty P[Z > t] dt$$

$$E[g(X,Y)] = \int_0^\infty P[g(X,Y) > t]dt$$

$$= \int_0^\infty \iint_{(x,y):g(x,y)>t} f_{XY}(x,y)dxdy dt$$

$$= \iint_{\mathbb{R}^2} \int_0^{g(x,y)} f_{XY}(x,y)dt dxdy$$

$$= \iint_{\mathbb{R}^2} g(x,y)f_{XY}(x,y)dxdy$$

Example 30.1: The positions $X \sim U(0, L)$ and $Y \sim U(0, L)$ of two persons on a road are independent. What is the mean distance between them?

Solution: Here,

$$f_{XY}(x,y) = \begin{cases} \frac{1}{L^2} & 0 < x < L, 0 < y < L \\ 0 & \text{else} \end{cases}$$

We want

$$E[|X - Y|] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |x - y| f_{XY}(x, y) dx dy$$
$$= \frac{1}{L^2} \int_0^L \int_0^L |x - y| dx dy$$

Now
$$\int_0^L |x-y| dx = \int_0^y (y-x) dx + \int_y^L (x-y) dx$$

$$=\frac{L^2}{2}+y^2-yL$$

So
$$E[|X - Y|] = \frac{1}{L^2} \int_0^L \frac{L^2}{2} + y^2 - yL \, dy$$

= $\frac{L}{3}$

Example 30.2: Show E[X + Y] = E[X] + E[Y].

Solution: [Continuous case only, discrete is similar]

With g(x, y) = x + y:

$$\begin{split} E[X+Y] &= E[g(X,Y)] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f_{XY}(x,y) dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x+y) f_{XY}(x,y) dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f_{XY}(x,y) dx dy + \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f_{XY}(x,y) dx dy \\ &= E[X] + E[Y] \end{split}$$

Note: by induction, $E[X_1 + \cdots + X_n] = E[X_1] + \cdots + E[X_n]$.

Example 30.3: Let X_1, X_2, \ldots, X_n be iid with (common) mean μ . The

quantity

$$\bar{X} = \frac{1}{n} \sum_{i=1}^{n} X_i$$

is called the **sample mean**. What is $E[\bar{X}]$?

Solution:

$$E[\bar{X}] = E\left[\frac{1}{n}\sum_{i=1}^{n} X_i\right]$$
$$= \frac{1}{n}E\left[\sum_{i=1}^{n} X_i\right]$$
$$= \frac{1}{n}\sum_{i=1}^{n} E[X_i]$$
$$= \frac{1}{n}\sum_{i=1}^{n} \mu$$
$$= \mu$$

Example 30.4: 10 friends play a game. Each has a ball and picks one of 10 targets randomly, independently of the others. Each has probability p of hitting their chosen target.

What is the expected number of targets not hit?

Solution:

Let $X_i = 1$ if target i is not hit, and 0 otherwise; X = # targets not hit.

Each person independently hits target i with probability p/10.

So
$$P[X_i = 1] = \left(1 - \frac{p}{10}\right)^{10}$$

 $E[X_i] = 1 \times P[X_i = 1] + 0 \times P[X_i = 0]$
 $= \left(1 - \frac{p}{10}\right)^{10}$

$$E[X] = E[X_1 + \dots + X_{10}]$$

$$= E[X_1] + \dots + E[X_{10}]$$

$$= 10 \left(1 - \frac{p}{10}\right)^{10}$$