

## Properties of Expectations

### Conditional Expectation [Ross S7.5]

**Example 34.1:** Recall that  $X$  and  $Y$  are jointly (bivariate) Gaussian (normal) with parameters:

$$\mu_X, \mu_Y, \sigma_X > 0, \sigma_Y > 0, -1 < \rho < 1$$

when  $f_{XY}(x, y)$  is given by

$$f_{XY}(x, y) = \frac{1}{2\pi\sigma_X\sigma_Y\sqrt{1-\rho^2}} \times \exp \left\{ -\frac{1}{2(1-\rho^2)} \left[ \left( \frac{x-\mu_X}{\sigma_X} \right)^2 + \left( \frac{y-\mu_Y}{\sigma_Y} \right)^2 - 2\rho \frac{(x-\mu_X)(y-\mu_Y)}{\sigma_X\sigma_Y} \right] \right\} \quad (28.1)$$

We now show that  $\rho$  is the correlation between  $X$  and  $Y$ .

From Notes #28:

$$\begin{aligned} E[X] &= \mu_X \\ E[Y] &= \mu_Y \\ Var[X] &= \sigma_X^2 \\ Var[Y] &= \sigma_Y^2 \end{aligned}$$

$$\text{Therefore } \rho(X, Y) = \frac{Cov[X, Y]}{\sigma_X\sigma_Y} = \frac{E[XY] - \mu_X\mu_Y}{\sigma_X\sigma_Y}$$

To determine  $E[XY]$ , recall from Notes #28 that  $f_{X|Y}(x|y)$  is Gaussian pdf where  $X$  has mean

$$\mu_X + \rho \frac{\sigma_X}{\sigma_Y} (y - \mu_Y).$$

So

$$E[X|Y = y] = \mu_X + \rho \frac{\sigma_X}{\sigma_Y} (y - \mu_Y)$$

$$\text{Now, } E[XY] = E[ E[XY|Y] ]$$

$$\begin{aligned} \text{and } E[XY|Y = y] &= E[Xy|Y = y] \\ &= yE[X|Y = y] \\ &= y \left( \mu_X + \rho \frac{\sigma_X}{\sigma_Y} (y - \mu_Y) \right) \\ &= \mu_X y + \rho \frac{\sigma_X}{\sigma_Y} (y^2 - \mu_Y y) \end{aligned}$$

$$\Rightarrow E[XY|Y] = \mu_X Y + \rho \frac{\sigma_X}{\sigma_Y} (Y^2 - \mu_Y Y)$$

$$\begin{aligned} \text{Therefore } E[XY] &= E[ E[XY|Y] ] \\ &= E \left[ \mu_X Y + \rho \frac{\sigma_X}{\sigma_Y} (Y^2 - \mu_Y Y) \right] \\ &= \mu_X E[Y] + \rho \frac{\sigma_X}{\sigma_Y} (E[Y^2] - \mu_Y E[Y]) \\ &= \mu_X \mu_Y + \rho \frac{\sigma_X}{\sigma_Y} (E[Y^2] - \mu_Y^2) \\ &= \mu_X \mu_Y + \rho \frac{\sigma_X}{\sigma_Y} Var[Y] \\ &= \mu_X \mu_Y + \rho \frac{\sigma_X}{\sigma_Y} \sigma_Y^2 \\ &= \mu_X \mu_Y + \rho \sigma_X \sigma_Y \end{aligned}$$

$$\begin{aligned} \Rightarrow \rho(X, Y) &= \frac{E[XY] - \mu_X \mu_Y}{\sigma_X \sigma_Y} \\ &= \frac{\rho \sigma_X \sigma_Y}{\sigma_X \sigma_Y} \\ &= \rho \end{aligned}$$

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### Computing Probabilities by Conditioning

We can use conditioning to compute probabilities:

Let  $A$  be an event.

Let random variable  $Y \in \{y_1, y_2, \dots\}$  and  $B_i = \{Y = y_i\}$ .

Then  $B_1, B_2, \dots$  partition the sample space  $S$ . So by law of total probability:

$$\begin{aligned} P[A] &= P[A|B_1]P[B_1] + P[A|B_2]P[B_2] + \dots \\ &= P[A|Y = y_1]P[Y = y_1] + P[A|Y = y_2]P[Y = y_2] + \dots \\ &= \sum_n P[A|Y = y_n]P[Y = y_n] \end{aligned}$$

Similarly, if  $Y$  is continuous:

$$P[A] = \int_{-\infty}^{\infty} P[A | Y = y] f_Y(y) dy$$

**Example 34.2:** Say  $X$  and  $Y$  are independent random variables with densities  $f_X(x)$  and  $f_Y(y)$ .

Find  $P[X < Y]$ .

*Solution:*

**Example 34.3:** Say  $X$  and  $Y$  are independent with densities  $f_X(x)$  and  $f_Y(y)$ . Find the cdf and pdf of  $X + Y$  by conditioning on  $Y$ .

*Solution:*