## **Properties of Expectations**

## **Conditional Expectation** [Ross S7.5]

Recall that for 2 discrete random variables X and Y with P[Y = y] > 0:

$$p_{X|Y}(x|y) = P[X=x|Y=y]$$
 
$$= \frac{p_{XY}(x,y)}{p_Y(y)}$$
 We can define the **conditional expectation**:

 $E[X|Y=y] = \sum_{x} x p_{X|Y}(x|y)$ 

Similarly, if 
$$X$$
 and  $Y$  are continuous, then provided  $f_Y(y) > 0$ :

 $f_{X|Y}(x|y) = \frac{f_{XY}(x,y)}{f_{Y}(y)},$ 

$$E[X|Y=y] = \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx$$

and

**Example 33.1:** Say 
$$X$$
 and  $Y$  have joint pdf [see Example 27.3]

 $f_{XY}(x,y) = \begin{cases} \frac{e^{-x/y}e^{-y}}{y} & 0 < x < \infty, \ 0 < y < \infty \\ 0 & \text{else} \end{cases}$ 

Find 
$$E[X|Y=y]$$
.   
Solution: From Example 27.3, for  $x>0, y>0$ 

 $f_{X|Y}(x|y) = \frac{f_{XY}(x,y)}{f_Y(y)}$ 

tion, e.g.,

and

So,

 $=\frac{1}{y}e^{-x/y}$ 

So, 
$$E[X|Y=y] = \int_0^\infty \frac{x}{y} e^{-x/y} dx = y$$

 $E[g(X)\mid Y=y] = \begin{cases} \sum_x g(x) p_{X\mid Y}(x|y) & \text{discrete case} \\ \\ \int_{-\infty}^{\infty} g(x) f_{X\mid Y}(x|y) dx & \text{continuous case} \end{cases}$ 

Note: Conditional expectations satisfy all the properties of ordinary expecta-

$$E\left[\sum_{i=1}^{n} X_i \mid Y = y\right] = \sum_{i=1}^{n} E[X_i | Y = y]$$

## E[X|Y = y] = yE[X|Y] = Y

E[X|Y=y] is a function of y, say g(y). Let E[X|Y] be g(Y), i.e., in Example 33.1:

**Computing Expectations by Conditioning** 

**Proposition 33.1** E[X] = E[E[X|Y]], *i.e.* 

 $E[X] = \sum_{y} E[X|Y = y]p_Y(y)$ 

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

Why? [Continuous Case] 
$$\int_{-\infty}^{\infty} E[X|Y=y] f_Y(y) dy = \int_{-\infty}^{\infty} \left[ \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx \right] f_Y(y) dy$$

[discrete case]

 $= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f_{X|Y}(x|y) f_Y(y) \ dxdy$  $= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f_{XY}(x,y) \ dxdy$ 

[continuous case]

E[X] = E[X|Y = 1]P[Y = 1]+E[X|Y=2]P[Y=2]+E[X|Y=3]P[Y=3]

 $\Rightarrow E[X] = 15$ 

with mean 50.

Solution: Let X = time to leave building, and Y = door choice.

 $= \frac{1}{3}(E[X|Y=1] + E[X|Y=2] + E[X|Y=3])$ 

Also, 
$$E[X|Y = 1] = 3$$
  
 $E[X|Y = 2] = 5 + E[X]$   
 $E[X|Y = 3] = 7 + E[X]$ 

 $E[X] = \frac{1}{3}(3+5+E[X]+7+E[X])$ 

Example 33.3: The number of people that enter a store in a day is random

The amount spent by each person is iid with mean \$8, and independent of the

number of people that enter. What is the expected amount spent in the store in one day? [Hard] 
$$Solution:$$
 Let  $N=\#$  customers that enter store in one day.

Let  $X_i$  = amount spent by ith customer.

Total amount spent is  $Y = \sum_{i=1}^{N} X_i$ .

 $E\left[\sum_{i=1}^{N} X_i\right] = E\left[E\left[\sum_{i=1}^{N} X_i \middle| N\right]\right]$ 

and  $E\left[\sum_{i=1}^{N} X_i \middle| N = n\right] = E\left[\sum_{i=1}^{n} X_i \middle| N = n\right]$  $= E \left[ \sum_{i=1}^{n} X_i \right]$ 

 $=\sum_{i=1}^{n}E\left[X_{i}\right]$ 

$$= nE[X_1]$$
 so 
$$E\left[\left.\sum_{i=1}^N X_i\right|N\right] = NE[X_1]$$

Therefore 
$$E\left[\sum_{i=1}^{N}X_i\right] = E\left[NE[X_1]\right]$$
  
=  $E[N]E[X_1]$   
=  $50 \times 8$