Math 3280 23-11-20  
Review.  
• Cov 
$$(X, Y) = E[(X-E[X])(Y-E[Y])]$$
  
 $= E[XY] - E[X] E[Y]$   
• Cov  $(X, Y) = 0$  if  $X, Y$  are independent.  
Prop 1. (1) Cov $(X, Y) = Cor(Y, X)$ .  
(2) Cov  $(X, X) = Var(X)$ .  
(3) Cov  $(aX, Y) = a Cov(X, Y)$ ,  $a \in \mathbb{R}$ .  
(4) Cov  $(\sum_{j=1}^{n} X_i, \sum_{j=1}^{m} Y_j)$   
 $= \sum_{i=1}^{n} \sum_{j=1}^{m} Cov(X_i, Y_j)$ .  
(1), (3), (4) imply that  $Cov(\cdot, \cdot)$  is bi-lenear.

Pf. Let us prove (4) only. Write 
$$\mu_{i} = E[X_{i}]$$
,  
 $V_{i} = E[Y_{j}]$ .  
Then by definition,  
 $Cov\left(\sum_{i=1}^{n} X_{i}, \sum_{j=1}^{m} Y_{j}\right) = E\left[\left(\sum_{i=1}^{n} X_{i} - \sum_{i=1}^{n} H_{i}\right) \cdot \left(\sum_{j=1}^{m} Y_{j} - \sum_{j=1}^{m} V_{j}\right)\right]$   
 $= E\left[\sum_{i=1}^{n} (X_{i} - H_{i}) \sum_{j=1}^{m} (Y_{j} - V_{j})\right]$   
 $= E\left[\sum_{i=1}^{n} \sum_{j=1}^{m} (X_{i} - H_{i}) (Y_{j} - V_{j})\right]$   
 $B_{j}$  the kinemity of  $E$   
 $= \sum_{i=1}^{n} \sum_{j=1}^{m} E\left[(X_{i} - H_{i}) (Y_{j} - V_{j})\right]$   
 $= \sum_{i=1}^{n} \sum_{j=1}^{m} C_{ov}(X_{i}, Y_{j})$ .  
Corollary 2.  $Var\left(\sum_{i=1}^{n} X_{i}\right) = \sum_{i=1}^{n} Var(X_{i}) + \sum_{i=1}^{2} Cov(X_{i}, X_{j})$ .  
Moreover if  $X_{1}, \cdots, X_{n}$  are pairwise independent,  
then  
 $Var\left(\sum_{i=1}^{n} X_{i}\right) = \sum_{i=1}^{n} Var(X_{i})$ .

§ 7.5 Conditional expectations.  
Def. If X and Y are discrete, then  
the conditional expectation of X given Y=Y, is  

$$E[X|Y=y] := \sum_{x} \times P\{X=x|Y=y\}$$
  
provided that  $P\{Y=y\} > 0$ .  
Def. In the case when X and Y are jointly  
cts with a density  $f(x,y)$ , the conditional  
expectation of X given Y=Y, is defined by  
 $E[X|Y=y] = \int_{-\infty}^{\infty} x \cdot f_{X|Y}(x|y) dx$ ,  
provided that  $f_{Y}(y) > 0$ , where  
 $f_{X|Y}(x|y) = \frac{f(x,y)}{f_{Y}(y)}$ .

Example 1. Let X, Y be jointly is with a density  

$$f(x, y) = \begin{cases} e^{-x/y} e^{-y}/y & \text{if } x, y > 0, \\ 0 & \text{otherwise}, \end{cases}$$
Calculate  $E[X|Y=y], y > 0.$   
Solution:  $f_{Y}(y) = \int_{-\infty}^{\infty} f(x, y) dx$   
 $= \int_{0}^{\infty} e^{-x/y} e^{-y}/y dx$   
 $= -e^{-x/y} e^{-y} \Big|_{x=0}^{\infty}$   
 $= e^{-y}, \text{ if } y > 0.$   
 $f_{X|Y}(x|y) = \frac{f(x, y)}{f_{Y}(y)}$ 

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

$$= \int_{0}^{\infty} x \cdot e^{-x/y} / y dx$$

$$I_{nt by pavt} x \cdot (-e^{-x/y}) \Big|_{x=0}^{+\infty} + \int_{0}^{\infty} e^{-x/y} dx$$

$$= 0 + (-y e^{-x/y}) \Big|_{X=0}^{+\infty}$$

$$= y \quad if \quad y > 0.$$

Now write  

$$E[X|Y] \text{ as a function of } Y \text{ by}$$

$$Y \longmapsto E[X|Y=y]$$

$$E[X|Y] \text{ is a r.v., the value of which depends on the value of Y.$$

$$Prop 2. E[X] = E[E[X|Y]]$$

$$Pf. \text{ We only prove it in the discrete care.}$$

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

$$= \sum_{Y} \sum_{X} x \cdot P\{X=x|Y=y\} \cdot P_{Y}(y)$$

$$= \sum_{Y} \sum_{X} x \cdot P\{X=x, Y=y\}$$

$$= \sum_{X} \sum_{Y} x \cdot P\{X=x, Y=y\}$$

$$= \sum_{X} x \cdot P\{X=x\} = E[X]$$

## Example 3.

A miner is trapped in a mine containing 3 doors. The first door leads to a tunnel that will take him to safety after 3 hours of travel. The second door leads to a tunnel that will return him to the mine after 5 hours of travel. The third door leads to a tunnel that will return him to the mine after 7 hours. If we assume that the miner is at all times equally likely to choose any one of the doors, what is the expected length of time until he reaches safety?



$$+ \in [X|Y=3] \cdot P\{Y=3\}$$

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

$$= \frac{1}{3} \left( 3 + (5 + \in [X]) + (7 + \in [X]) \right)$$
Soluting this equation, we obtain
$$E[X] = 3 + 5 + 7 = 15 \quad (hours)$$