

1	2	3	4	Total

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## KOM505E - Probability Theory and Stochastic Processes Midterm #2

Dec. 10, 2015

**Rules:**

- Closed book & notes.
- Cell phones are not allowed.
- This exam will count for 20% of your final grade.
- Duration: 110 min.

1. (25 pts) Consider two random variables  $X$  and  $Y$ . Prove that if  $Y = aX + b$  where  $a$  and  $b$  are constants, then the correlation coefficient  $\rho$  is given as

$$\rho_{X,Y} = \begin{cases} 1 & \text{if } a > 0 \\ -1 & \text{if } a < 0 \end{cases}$$

$$\rho_{X,Y} = \frac{\text{Cov}(X,Y)}{\sigma_X \cdot \sigma_Y}$$

$$\text{Cov}(X,Y) = E\{(X - \mu_X)(Y - \mu_Y)\}$$

$$\begin{aligned} \mu_Y &= E\{aX + b\} = a E\{X\} + b \\ &= a\mu_X + b \end{aligned}$$

then

$$\begin{aligned} \text{Cov}(X,Y) &= E\left\{(X - \mu_X) \underbrace{(aX + b - a\mu_X - b)}_{a(X - \mu_X)}\right\} \\ &= E\{a(X - \mu_X)^2\} = a\sigma_X^2 \end{aligned}$$

In addition:

$$\sigma_Y^2 = a^2 \sigma_X^2 \Rightarrow \sigma_Y = |a| \sigma_X$$

Then

$$\rho_{X,Y} = \frac{a\sigma_X^2}{\sigma_X \cdot |a| \sigma_X} = \frac{a}{|a|} = \text{sgn}(a) = \begin{cases} +1 & \text{if } a > 0 \\ -1 & \text{if } a < 0 \end{cases}$$

2. (25 pts) If a joint pdf is given as  $p_{X,Y}(x,y) = (\frac{1}{4})^2 \exp\{-\frac{1}{2}(|x| + |y|)\}$  for  $-\infty < x, y < \infty$ . New random variables  $W$  and  $Z$  are obtained from  $X$  and  $Y$  as follows:

$$\begin{bmatrix} W \\ Z \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} X \\ Y \end{bmatrix}$$

- (a) What is the range of  $W$  and  $Z$ ?  
 (b) Find the joint pdf of  $W$  and  $Z$ .

a) As  $\begin{cases} -\infty < x < \infty \\ -\infty < y < \infty \end{cases} \Rightarrow \begin{cases} W = 2X + 1Y \Rightarrow -\infty < W < \infty \\ Z = 3X + 2Y \Rightarrow -\infty < Z < \infty \end{cases}$

b)  $G = \begin{bmatrix} 2 & 1 \\ 3 & 2 \end{bmatrix} \Rightarrow |G| = |G^{-1}| = 1$   
 $G^{-1} = \begin{bmatrix} 2 & -1 \\ -3 & 2 \end{bmatrix} \Rightarrow \begin{bmatrix} X \\ Y \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ -3 & 2 \end{bmatrix} \begin{bmatrix} W \\ Z \end{bmatrix}$

Hence

$$f_{W,Z}(w,z) = \left(\frac{1}{4}\right)^2 \exp\left\{-\frac{1}{2}(|2W-Z| + |-3W+2Z|)\right\} \quad -\infty < w, z < \infty$$

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3. (25 pts) Consider two discrete random variables  $X$  and  $Y$ , whose joint PMF are given below:

	$Y = -8$	$Y = 0$	$Y = 8$
$X = -6$	$1/12$	$1/6$	$1/12$
$X = 0$	$1/6$	$1/12$	$1/12$
$X = 6$	$1/12$	$1/12$	$K$

- Find the value of  $K$ .
- Find the conditional probability of  $P(X = 0|Y = 8)$ .
- Are  $X$  and  $Y$  independent? Show your work.
- Find the expected value of  $X$  and  $Y$ .
- Find the covariance matrix  $C$  of  $(X, Y)$ .
- Now, assuming that the covariance matrix of  $X$  and  $Y$  are given by (for the sake of easy computations):

$$C = \begin{bmatrix} 6 & 2 \\ 2 & 3 \end{bmatrix},$$

find a matrix  $A$  that transforms  $X$  and  $Y$  to uncorrelated random variables  $W$  and  $Z$ :

$$\begin{bmatrix} W \\ Z \end{bmatrix} = A \begin{bmatrix} X \\ Y \end{bmatrix}$$

a) The values in the table should add up to 1.

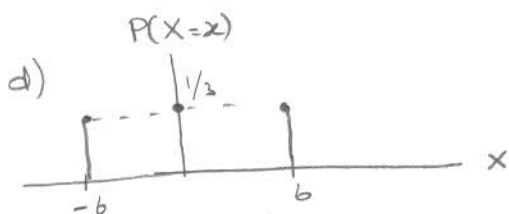
$$\text{Hence } K = 1/6$$

$$b) P(X=0|Y=8) = \frac{P(X=0, Y=8)}{P(Y=8)} = \frac{1/12}{1/3} = 1/4$$

$$P(Y=8) = \sum_{k \in \{-6, 0, 6\}} P(X=k, Y=8) = 1/12 + 1/12 + 1/6 = 1/3$$

$$c) P(X=0) = \sum_{k \in \{-8, 0, 8\}} P(X=0, Y=k) = 1/6 + 1/12 + 1/12 = 1/3$$

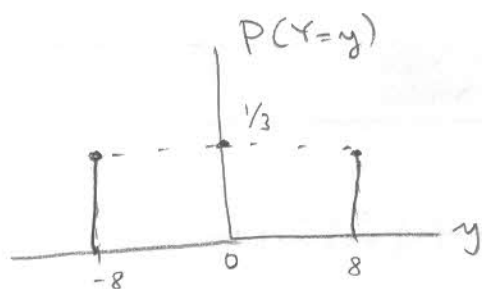
$P(X=0|Y=8) \neq P(X=0) \Rightarrow X$  &  $Y$  are NOT independent.



$$\Rightarrow E[X] = \frac{1}{3}(-6) + \frac{1}{3}(0) + \frac{1}{3}(6) = 0$$

$$\mu_X = 0$$

Similarly



$$E[Y] = \frac{1}{3}(-8) + \frac{1}{3}(0) + \frac{1}{3}(8)$$

$$\mu_Y = 0$$

$$e) \text{Cov}(X, Y) = E((X - \mu_X)(Y - \mu_Y)) = E[XY]$$

$$= \frac{1}{12}(-6 \cdot -8) + \frac{1}{12}(-6 \cdot 8) + \frac{1}{12}(6 \cdot -8) + \frac{1}{12}(6 \cdot 8)$$

$$= +4 - 4 - 4 + 4 = 0$$

$$\sigma_X^2 = E[X^2] = \frac{1}{3}(-6)^2 + \frac{1}{3}0^2 + \frac{1}{3}6^2 = 24$$

$$\sigma_Y^2 = E[Y^2] = \frac{1}{3}(-8)^2 + \frac{1}{3}0^2 + \frac{1}{3}8^2 = 128/3$$

Then

$$C = \begin{bmatrix} 24 & 0 \\ 0 & 128/3 \end{bmatrix}$$

$$f) \quad C = \begin{bmatrix} 6 & 2 \\ 2 & 3 \end{bmatrix} \quad \det(\lambda I - C) = \begin{vmatrix} 6-\lambda & 2 \\ 2 & 3-\lambda \end{vmatrix} = (6-\lambda)(3-\lambda) - 4 = 0$$

$$= 18 - 9\lambda + \lambda^2 - 4 = 0$$

$$(\lambda - 2)(\lambda - 7) = 0$$

$\lambda = 2$   
 $\lambda = 7$  } eigenvalues of C

$$\text{For } \lambda = 2 \quad \begin{bmatrix} 6 & 2 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = 2 \begin{bmatrix} a \\ b \end{bmatrix}$$

$$6a + 2b = 2a$$

$$b = -2a$$

$$v_1 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$\lambda = 7 \quad \begin{bmatrix} 6 & 2 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = 7 \begin{bmatrix} a \\ b \end{bmatrix}$$

$$6a + 2b = 7a$$

$$2b = a$$

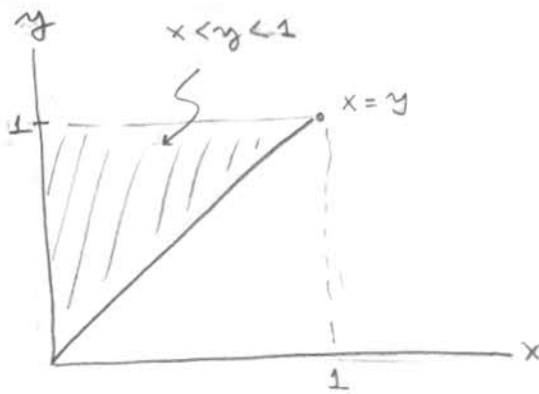
$$v_2 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\text{Hence} \quad A = \begin{bmatrix} 1/\sqrt{5} & 2/\sqrt{5} \\ -2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix}$$

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4. (25 pts) The joint pdf  $p_{X,Y}$  of two random variables  $X$  and  $Y$  are given as uniform over the region  $0 < y < 1$  and  $0 < x < y$ , and zero otherwise.

- (a) Find the conditional PDF  $p_{Y|X}(y|x)$ .  
 (b) Find the minimum mean square error (MMSE) estimate  $\hat{Y}$  given  $X = x$ , assuming a nonlinear prediction model.  
 (c) Plot the estimated value of  $\hat{Y}$  versus  $x$ , by also indicating the region in the  $x-y$  plane for which the joint PDF is nonzero.  
 (d) Based on the plot of the predictor, comment on the intuition behind this predictor.



$$f_{X,Y}(x,y) = \begin{cases} 2 & \text{in the shaded area} \\ 0 & \text{otherwise} \end{cases}$$

a)  $f_X(x) = \int_x^1 f_{X,Y}(x,y) dy = 2y \Big|_x^1 = 2(1-x) \quad 0 < x < y$

$$f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)} = \frac{2}{2(1-x)} = \frac{1}{1-x} \quad \begin{matrix} 0 < x < y \\ x < y < 1 \end{matrix}$$

b)  $\hat{Y} = E_{Y|X}[Y|X] = \int_x^1 y \cdot f_{Y|X}(y|x) dy$   

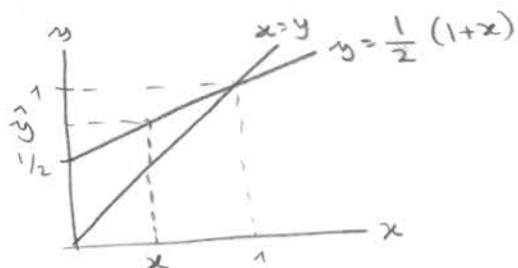
$$= \int_x^1 \frac{y}{1-x} dy = \frac{1}{1-x} \cdot \frac{y^2}{2} \Big|_x^1 = \frac{1}{2} \left( \frac{1}{1-x} \cdot (1-x^2) \right)$$
  

$$= \frac{1}{2} (1+x)$$

c)

Hence

$$\hat{Y} = \frac{1}{2} (1+x)$$



d) Given an  $x$  value, the predictor  $\hat{y}$  picks the average of all possible  $y$  values.

$$\hat{y} = E_{y|x}(y|x)$$