

# ECE 316-Solutions of Problem Set 5

March 9, 2011

## Solution 1

Our estimate of  $X$  is  $\hat{X} = aY + b$  where  $a$  and  $b$  should be chosen to minimize the mean squared error which is  $J = E(X - \hat{X})^2$ . To minimize  $J$ , we set  $\frac{\partial J}{\partial a} = 0$  and  $\frac{\partial J}{\partial b} = 0$

Now:

$$J = E(X - aY - b)^2 = E[X^2] - 2aE[XY] - 2bE[X] + 2abE[Y] + a^2E[Y^2] + b^2$$

Therefore,

$$\begin{aligned}\frac{\partial J}{\partial a} &= 0 \\ \Rightarrow E(2(X - (aY + b))(-Y)) &= 0 \\ \Rightarrow E(XY) &= aE(Y^2) + bE(Y)\end{aligned}\tag{1}$$

$$\begin{aligned}\frac{\partial J}{\partial b} &= 0 \\ \Rightarrow E((2(X - (aY + b))(-1)) &= 0 \\ \Rightarrow E(X) &= aE(Y) + b\end{aligned}\tag{2}$$

Solving 1 and 2 for  $a$  and  $b$

$$\begin{aligned}a &= \frac{E(XY) - E(X)E(Y)}{E(Y^2) - E(Y)^2} \\ &= \frac{\text{cov}(X, Y)}{\text{Var}(Y)}\end{aligned}$$

and

$$b = E(X) - aE(Y)$$

## Problem 2

$$\begin{aligned}f_X(z) &= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{z^2}{2}\right) \quad -\infty < z < \infty \\E(Z) &= E[X \mathbf{1}_{[X>x]}] = \int_x^\infty z \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{z^2}{2}\right) dz \\&= -\frac{1}{\sqrt{2\pi}} \exp\left(-\frac{z^2}{2}\right) \Big|_x^\infty \\&= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right)\end{aligned}$$

## Problem 3

This is an important result.

First note by definition:

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy = h(x) \int_{-\infty}^{\infty} g(y) dy = C_1 h(x)$$

where  $C_1 = \int_{-\infty}^{\infty} g(y) dy$

Similarly, we obtain

$$f_Y(y) = C_2 g(y)$$

where  $C_2 = \int_{-\infty}^{\infty} h(x) dx$ .

On the other hand we know

$$1 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx dy = \int_{-\infty}^{\infty} h(x) dx \int_{-\infty}^{\infty} g(y) dy = C_2 C_1$$

Therefore  $f_{X,Y}(x,y) = f_X(x)f_Y(y)$  or the random variables are independent.

Now consider the joint density given in the example, then it can be written as:

$$f_{X,Y}(x,y) = Cxy \mathbf{1}_{[0<x<1]} \mathbf{1}_{[0<y<1]} \mathbf{1}_{[0<x+y<1]}, \quad -\infty < x < \infty; -\infty < y < \infty$$

which clearly cannot be written as  $h(x)g(y)$  for some functions  $h(\cdot)$  and  $g(\cdot)$  because the last indicator function introduces dependency because it involves both  $x$  and  $y$ .

Therefore  $f_{X,Y}(x,y) \neq f_X(x)f_Y(y)$  and so the random variables cannot be independent.

## Problem 4

Instead of solving this problem for three random variables we can solve it for the case of  $n$  independent random variables.

$X_1, \dots, X_n$  are i.i.d r.v having uniform distribution over  $(0, 1)$  Let  $Y = \max(X_1, \dots, X_n)$ ,  $Z = \min(X_1, \dots, X_n)$

$P(Y \leq y) = P(X_1 \leq y, X_2 \leq y, \dots, X_n \leq y) = P(X_1 \leq y)P(X_2 \leq y) \cdots P(X_n \leq y)$  by independence

$$= \begin{cases} 0, & y < 0 \\ y^n, & 0 < y < 1 \\ 1, & y > 1 \end{cases}$$

$$f_Y(y) = ny^{n-1} \quad 0 < y < 1$$

$$P(Z \leq z) = 1 - P(Z > z) \\ = 1 - P(X_1 > z)P(X_2 > z) \cdots P(X_n > z)$$

$$= \begin{cases} 1, & y < 0 \\ 1 - (1 - z)^n, & 0 < z < 1 \\ 0, & z > 1 \end{cases}$$

$$f_Z(z) = n(1 - z)^{n-1} \quad 0 < z < 1$$

$$E(Y) = \frac{n}{n+1}$$

$$E(Z) = \frac{1}{n+1}$$

So we see that as  $n \rightarrow \infty$   $E[Y] \rightarrow 1$  and  $E[Z] \rightarrow 0$  i.e. as the number of random variables increases the largest value we obtain is closer and closer to 1 the maximum a given r.v can be and the minimum goes towards 0 which is the minimum value a given r.v. can have.

## Problem 5

$$\begin{aligned} E(X) &= \int_{y=0}^{\infty} \int_{x=0}^{\infty} x \frac{1}{y} \exp(-(y + x/y)) dx dy \\ &= \int_{y=0}^{\infty} \frac{1}{y} \exp(-y) \int_{x=0}^{\infty} x \exp(-x/y) dx dy \\ &= \int_{y=0}^{\infty} \frac{1}{y} \exp(-y) y^2 dy \\ &= 1 \end{aligned}$$

$$\begin{aligned} E(Y) &= \int_{y=0}^{\infty} \int_{x=0}^{\infty} y \frac{1}{y} \exp(-(y + x/y)) dx dy \\ &= 1 \end{aligned}$$

$$\begin{aligned} E(XY) &= \int_{y=0}^{\infty} \int_{x=0}^{\infty} xy \frac{1}{y} \exp(-(y + x/y)) dx dy \\ &= \int_{y=0}^{\infty} y \frac{1}{y} \exp(-y) \int_{x=0}^{\infty} x \exp(-x/y) dx dy \\ &= \int_{y=0}^{\infty} \exp(-y) y^2 dy \\ &= 2 \end{aligned}$$

$$\text{Cov}(X, Y) = E(XY) - E(X)E(Y) = 1$$

## Problem 6

$X_1, X_2, X_3, X_4$  are pairwise uncorrelated random variables each having mean 0 and variance 1.

$$\begin{aligned} \text{cov}(X_1 + X_2, X_2 + X_3) &= \text{cov}(X_1, X_2) + \text{cov}(X_1, X_3) + \text{cov}(X_2, X_2) + \text{cov}(X_2, X_3) \\ &= 0 + 0 + 1 + 0 = 1 \end{aligned}$$

$$\begin{aligned} \text{var}(X_1 + X_2) &= \text{var}(X_1) + \text{var}(X_2) + 2\text{cov}(X_1, X_2) \\ &= 1 + 1 + 0 = 2 \end{aligned}$$

$$\begin{aligned} \text{var}(X_2 + X_3) &= \text{var}(X_2) + \text{var}(X_3) + 2\text{cov}(X_2, X_3) \\ &= 1 + 1 + 0 = 2 \end{aligned}$$

$$\begin{aligned} \rho(X_1 + X_2, X_2 + X_3) &= \frac{\text{cov}(X_1 + X_2, X_2 + X_3)}{\sqrt{\text{var}(X_1 + X_2)\text{var}(X_2 + X_3)}} \\ &= 0.5 \end{aligned}$$

$$\begin{aligned} \text{cov}(X_1 + X_2, X_3 + X_4) &= \text{cov}(X_1, X_3) + \text{cov}(X_1, X_4) + \text{cov}(X_2, X_3) + \text{cov}(X_2, X_4) \\ &= 0 \end{aligned}$$

$$\rho(X_1 + X_2, X_3 + X_4) = 0$$

## Problem 7

$$\begin{aligned}f(x, y) &= \frac{e^{-x/y}e^{-y}}{y} & 0 < x < \infty, 0 < y < \infty \\f_Y(y) &= \int_{x=0}^{\infty} \frac{e^{-x/y}e^{-y}}{y} dx \\&= e^{-y} & y > 0 \\f_{X|Y}(x|y) &= \frac{f(x, y)}{f(y)} = \frac{e^{-x/y}}{y} \\E(X^2|Y = y) &= \int_{x=0}^{\infty} x^2 \frac{e^{-x/y}}{y} dx \\&= 2y^2\end{aligned}$$

## Problem 8

This is the same argument as in Problem 1 with the linear term in  $Y$  missing.

Let  $J = E((X - a)^2)$ . To minimize  $J$ , we should have  $\frac{dJ}{da} = 0$ . Therefore

$$\begin{aligned}\frac{dJ}{da} &= 0 \\ \Rightarrow 2E(X - a) &= 0 \\ \Rightarrow a &= E(X)\end{aligned}$$

The second part needs some proof.

First note that  $|X - a| = (X - a)\mathbf{1}_{[X > a]} - (X - a)\mathbf{1}_{[X \leq a]}$ .

Therefore:

$$E|X - a| = \int_a^{\infty} (x - a)f_X(x)dx - \int_{-\infty}^a (x - a)f_X(x)dx$$

Differentiating (using Leibniz's rule) w.r.t  $a$  and setting the derivative to 0 gives:

$$-\int_a^{\infty} f_X(x)dx + \int_{-\infty}^a f_X(x)dx = 0$$

Noting that  $\int_{-\infty}^a f_X(x)dx = F_X(a)$  and  $\int_a^{\infty} f_X(x)dx = 1 - F_X(a)$  we obtain:

$$F_X(a) = 1 - F_X(a)$$

or  $F_X(a) = 0.5$  so  $a$  corresponds to the median.

## Problem 9

Without loss of generality assume  $E[X] = E[Y] = 0$  then

$$\text{Cov}(X + Y, X - Y) = E[(X - Y)(X + Y)] = E[X^2] - E[Y^2] = (\text{Var}(X) - \text{Var}(Y)) = 0$$

since  $X$  and  $Y$  are identically distributed.

## Problem 10

Without loss of generality assume  $E[X] = E[Y] = 0$ .

Assume  $E[Y^2] > 0$  and let  $a = \frac{E[XY]}{E[Y^2]}$ .

Then

$$\begin{aligned} 0 &\leq E[Y^2]E[(X - aY)^2] \\ &= E[Y^2](E[X^2] + a^2E[Y^2] - 2aE[XY]) \\ &= E[Y^2]E[X^2] - (E[XY])^2 \end{aligned}$$

from which the result follows.

## Problem 11

Once again without loss of generality assume that  $E[X] = E[Y] = 0$  (convince yourselves that if  $E[X] = m_X$  and  $E[Y] = m_Y$  the answer still holds. Now  $Y = a + bX$ ,

Hence  $cov(XY) = E(XY) = aE[X] + bE[X^2] = bvar(X)$

Now  $var(Y) = b^2Var(X)$

Therefore

$$\rho(X, Y) = \frac{bvar(X)}{|b|var(X)} = sign(b)$$

from which the answer follows.