

# Problem Set 3

Daniel Shapiro

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## Question 1 Background:

Suppose that  $X$  and  $Y$  are identical and independently distributed (i.i.d.) random variables distributed  $\mathcal{N}(\mu_X, \sigma_X^2)$  and  $\mathcal{N}(\mu_Y, \sigma_Y^2)$ , respectively. Note that  $\sigma_X$  represents the standard deviation of  $X$ , and  $\sigma_X^2$  the variance. Find the following, expressed in terms of  $\mu$  and  $\sigma$ :

1a)  $E(7X - 6Y + 12)$

Additivity and homogeneity are two properties of expectations. Thus, we can write this expression as  $7E[X] - 6E[Y] + 12$ , or  $7\mu_X - 6\mu_Y + 12$ .

1b)  $\text{Var}(X + 5Y)$

We know that  $\text{Var}(aX + bY + c) = a^2\text{Var}(X) + b^2\text{Var}(Y) + 2ab\text{Cov}(X, Y)$ . Inserting a 1 and 5 as a and b, we get  $\sigma_X^2 + 25\sigma_Y^2 + 10\text{Cov}(X, Y)$ .  $\text{Cov}(X, Y) = E[XY] - E[X]E[Y] = E[(X - \mu_X)(Y - \mu_Y)]$ . So then we get:  $\text{Var}(X + 5Y) = \sigma_X^2 + 25\sigma_Y^2 + 10E[(X - \mu_X)(Y - \mu_Y)]$ .

Because we know that the two variables are so-called “i.i.d.” variables, we know that the co-variance is 0. Thus, we can remove the last term entirely and end up with  $\text{Var}(X + 5Y) = \sigma_X^2 + 25\sigma_Y^2$ .

1c)  $E(5X^2 - 12XY + 16Y^2)$

First, we can split these up according to properties shown in 1a. So we get  $5E[X^2] - 12E[XY] + 16E[Y^2]$ .

Now, we can already deal with the first and last bits of this expression. We know that  $\sigma_X^2 = E[X^2] - \mu_X^2$  and  $\sigma_Y^2 = E[Y^2] - \mu_Y^2$ . Thus,  $E[X^2] = \sigma_X^2 + \mu_X^2$  and  $E[Y^2] = \sigma_Y^2 + \mu_Y^2$ . So we can set this as  $5\sigma_X^2 + 5\mu_X^2 + 12E[XY] + 16\sigma_Y^2 + 16\mu_Y^2$ .

Finally, the question tells us that  $X$  and  $Y$  are independent random variables. So we know therefore that  $E[XY] = E[X]E[Y]$ . We can write this as  $\mu_X\mu_Y$ .

Our final expression:  $5\sigma_X^2 + 5\mu_X^2 + 12\mu_X\mu_Y + 16\sigma_Y^2 + 16\mu_Y^2$ .

## Question 2 Background:

Papers like *this one* have found that married couples tend to sort along ideological lines (“assortative mating”). Say we know the true distribution of ideology among heterosexual married couples along a scale of 1 (very conservative) to 4 (very liberal). The joint distribution for  $X$  (the woman’s ideology) and  $Y$  (the man’s ideology) is:

| Y | X   |     |     |     |
|---|-----|-----|-----|-----|
|   | 1   | 2   | 3   | 4   |
| 1 | .12 | .05 | .01 | .01 |
| 2 | .08 | .17 | .05 | .02 |
| 3 | .02 | .01 | .23 | .05 |
| 4 | .02 | .02 | .04 | .10 |

## 2a) What is the expected value of $X$ and the expected value of $Y$ ?

To find the expected value of  $X$ , we can run the formula:  $E[X] = \sum(x)P(x)$ . So we multiply the value  $x$  by each corresponding probability and add them up. So  $E[X] = ((1 * .12) + (1 * .08) + ... + (4 * .10))$ . The final answer is **2.45**.

To find the expected value of  $Y$ , we can run the same formula, but for  $Y$  instead of  $X$ . So we multiply the value  $y$  by each corresponding probability and add them up. So  $E[Y] = ((1 * .12) + (1 * .05) + ... + (4 * .10))$ . The final answer is **2.48**.

## 2b) What are the variances of $X$ and of $Y$ ?

The variance of  $X$  or ( $Var(X)$ ) can be defined as  $\sum(x - \mu_x)^2 p(x)$ . The variance of  $Y$  or ( $Var(Y)$ ) can be defined as  $\sum(y - \mu_y)^2 p(y)$ . So we already know that  $\mu_x = 2.45$  and  $\mu_y = 2.48$ ; now, we need to find the probability of  $X$  at each point and the probability of  $Y$  at each point. To do this, we can just add up the proper rows and columns.

For  $X$ , we get: (.24 (when  $x = 1$ ), .25 (when  $x = 2$ ), .33 (when  $x = 3$ ), and .18 (when  $x = 4$ )).

For  $Y$ , we get: (.19 (when  $y = 1$ ), .32 (when  $y = 2$ ), .31 (when  $y = 3$ ), and .18 (when  $y = 4$ )).

Now, equation time:

$Var(X) = (1 - 2.45)^2 * .24 + (2 - 2.45)^2 * .25 + (3 - 2.45)^2 * .33 + (4 - 2.45)^2 * .18$ . This ends up equaling **1.0875**.

$Var(Y) = (1 - 2.48)^2 * .19 + (2 - 2.48)^2 * .32 + (3 - 2.48)^2 * .31 + (4 - 2.48)^2 * .18$ . This ends up equaling **0.9896**.

## 2c) What is the covariance of $X$ and $Y$ ?

We know that the covariance of  $X$  and  $Y$  can be defined by the function:

$$Cov(X, Y) = E[XY] - E[X]E[Y]$$

First, we know  $E[X]E[Y]$  – it's  $2.48 * 2.45$ . This is 6.076.

To find  $E[XY]$ , we need to multiply each value of  $Y$  and  $X$  by their probability, and then add everything together. Like so:

$$((1 * 1 * .12) + (1 * 2 * .08) + (1 * 3 * .03) + ... + (4 * 4 * .10)).$$

After a somewhat time-consuming process, we end up with 6.70.

Then, we subtract:  $6.70 - 6.076 = \mathbf{0.624}$ .

## 2d) What is the correlation between $X$ and $Y$ ?

The function for correlation is:

$$Cor[X, Y] = \frac{Cov[X, Y]}{\sqrt{V[X]V[Y]}}.$$

Luckily, we know all of these values already. Plugging in our numbers, we get:

$$\text{Cor}[X, Y] = \frac{0.624}{\sqrt{1.0875 * 0.9896}} = \mathbf{0.6015}.$$

**2e) Find  $E[Y|X = x]$  for  $x = 1, 2, 3, 4$ . Find the probability mass function of the random variable  $E[Y|X]$ .**

For purposes of this question, we will set  $E[X|Y] = Z$ . So for this question, we need to take each value of  $X=x$  and multiply it by the given probability value divided by the marginal probability of  $Y=y$ . For  $E[Y|X = 1]$ , we thus get the following:

$$((1 * \frac{.12}{.24}) + (2 * \frac{.08}{.24}) + (3 * \frac{.02}{.24}) + (4 * \frac{.02}{.24})) = 1.75$$

For  $E[Y|X = 2]$ , we get:

$$((1 * \frac{.05}{.25}) + (2 * \frac{.17}{.25}) + (3 * \frac{.01}{.25}) + (4 * \frac{.02}{.25})) = 2$$

For  $E[Y|X = 3]$ , we get:

$$((1 * \frac{.01}{.33}) + (2 * \frac{.05}{.33}) + (3 * \frac{.23}{.33}) + (4 * \frac{.04}{.33})) = 2.91$$

For  $E[Y|X = 4]$ , we get:

$$((1 * \frac{.01}{.18}) + (2 * \frac{.02}{.18}) + (3 * \frac{.05}{.18}) + (4 * \frac{.10}{.18})) = 3.33$$

Writing out the probability mass function is relatively easy if we've already figured out  $Z$  at  $x = 1, 2, 3, 4$ . Below:

$$p(z) = \begin{cases} .24 & \text{if } Z = 1.75 \\ .25 & \text{if } Z = 2 \\ .33 & \text{if } Z = 2.91 \\ .18 & \text{if } Z = 3.33 \end{cases}$$

**2f) Find  $\text{Var}(Y|X)$ .**

We know:

$$\text{Var}(Y|X) = \begin{cases} \text{Var}(Y|X = 1) & \text{if } X = 1 \\ \text{Var}(Y|X = 2) & \text{if } X = 2 \\ \text{Var}(Y|X = 3) & \text{if } X = 3 \\ \text{Var}(Y|X = 4) & \text{if } X = 4 \end{cases}$$

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$$\text{Var}(Y|X) = \begin{cases} \text{Var}(Y|X = 1) & \text{w/prob of .24} \\ \text{Var}(Y|X = 2) & \text{w/prob of .25} \\ \text{Var}(Y|X = 3) & \text{w/prob of .33} \\ \text{Var}(Y|X = 4) & \text{w/prob of .18} \end{cases}$$

$$\text{Var}(Y|X = 1) = E[Y^2|X = 1] - E[Y|X = 1]^2$$

$$= ((1^2 * \frac{.12}{.24}) + (2^2 * \frac{.08}{.24}) + (3^2 * \frac{.02}{.24}) + (4^2 * \frac{.02}{.24})) - 1.75^2 = \frac{47}{12} - 3.0625 = 0.854166.$$

$$\text{Var}(Y|X = 2) = E[Y^2|X = 2] - E[Y|X = 2]^2$$

$$= ((1^2 * \frac{.05}{.25}) + (2^2 * \frac{.17}{.25}) + (3^2 * \frac{.01}{.25}) + (4^2 * \frac{.02}{.25})) - 2^2 = \frac{114}{25} - 4 = 0.56.$$

$$\begin{aligned} Var(Y|X = 3) &= E[Y^2|X = 3] - E[Y|X = 3]^2 \\ &= ((1^2 * \frac{.01}{.33}) + (2^2 * \frac{.05}{.33}) + (3^2 * \frac{.23}{.33}) + (4^2 * \frac{.04}{.33})) - 2.91^2 = \frac{292}{33} - 8.4628 = 0.3857. \end{aligned}$$

$$\begin{aligned} Var(Y|X = 4) &= E[Y^2|X = 4] - E[Y|X = 4]^2 \\ &= ((1^2 * \frac{.01}{.18}) + (2^2 * \frac{.02}{.18}) + (3^2 * \frac{.05}{.18}) + (4^2 * \frac{.10}{.18})) - 3.33^2 = \frac{214}{18} - 11.1 = 0.78. \end{aligned}$$

**So, this can be expressed as:**

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$$Var(Y|X) = \begin{cases} 0.854166 & \text{w/prob of .24} \\ 0.56 & \text{w/prob of .25} \\ 0.3857 & \text{w/prob of .33} \\ 0.78 & \text{w/prob of .18} \end{cases}$$