

# Problem Set 3

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

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

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

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

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

$$\begin{aligned}
\text{Var}(Y|X=3) &= E[Y^2|X=3] - E[Y|X=3]^2 \\
&= ((1^2 * \frac{.01}{.33}) + (2^2 * \frac{.05}{.33}) + (3^2 * \frac{.23}{.33}) + (4^2 * \frac{.04}{.33})) - 2.91^2 = \frac{292}{33} - 8.4628 = 0.3857. \\
\text{Var}(Y|X=4) &= E[Y^2|X=4] - E[Y|X=4]^2 \\
&= ((1^2 * \frac{.01}{.18}) + (2^2 * \frac{.02}{.18}) + (3^2 * \frac{.05}{.18}) + (4^2 * \frac{.10}{.18})) - 3.33^2 = \frac{214}{18} - 11.1 = 0.78.
\end{aligned}$$

So, this can be expressed as:

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

**2g) Show that  $E[Y] = E[E[Y|X]]$ .**

We already know that  $E[Y] = 2.48$ . Now, to get  $E[E[Y|X]]$ , we multiply each value of  $E[Y|X]$  (which we called Z) by its corresponding probability and add them together. We thus end up with:

$$1.75 * .24 + 2 * .25 + 2.91 * .33 + 3.33 * .18.$$

Added together, these equal 2.4797. While 2.4797 is *technically* .0003 off from 2.48, this is almost certainly due to my rounding throughout the previous few problems. Thus, we can say that 2.4797 is essentially the same as 2.48 here, and thereby confidently say:  $2.48 = 2.48$ ; *thus*,  $E[Y] = E[E[Y|X]]$ .

**2h) Find  $\text{Var}(Y)$  using the law of total variance. Explain in words what the law of total variance does.**

The law of total variance can be expressed as:  $V[Y] = E[V[Y|X]] + V[E[Y|X]]$ .