

# Math 170A Homework 3

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## Basics

### Question 1

Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space and let  $(E_0, \mathcal{E}_0)$  and  $(E_1, \mathcal{E}_1)$  be finite, discrete measurable spaces. Let  $X : \Omega \rightarrow E_0$  be a measurable function, so that  $(E_0, \mathcal{E}_0, \mathbb{P}_X)$  is a probability space. Let  $h : E_0 \rightarrow E_1$  be an arbitrary function (which is necessarily measurable since  $\mathcal{E}_0 = \mathcal{P}(E_0)$ ), and let  $Y = h \circ X : \Omega \rightarrow E_1$  be the composition (which is thus measurable).  
Prove that the measure  $\mathbb{P}_1$  on  $E_1$  induced by  $Y : \Omega \rightarrow E_1$ , when viewed as an  $E_1$ -valued random variable on the probability space  $(\Omega, \mathcal{F}, P)$ , is the same as the measure  $\tilde{\mathbb{P}}_1$  on  $E_1$  induced by  $h : E_0 \rightarrow E_1$ , when viewed as an  $E_1$ -valued random variable on the probability space  $(E_0, \mathcal{E}_0, \mathbb{P}_X)$ .

We want to show that  $\mathbb{P}_1 = \tilde{\mathbb{P}}_1$ . By definition, for  $x \in E_1$

$$\mathbb{P}_1(\{x\}) = \mathbb{P}(Y^{-1}(\{x\}))$$

which is the probability measure of the preimage in the probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ . By the definition of  $Y$ , we then have that

$$\mathbb{P}(Y^{-1}(\{x\})) = \mathbb{P}(X^{-1} \circ h^{-1}(\{x\}))$$

But  $h^{-1}(\{x\})$  may contain more than one point in  $E_0$ , thus, we can say that

$$X^{-1}(h^{-1}(\{x\})) = \bigcup_{y \in h^{-1}(\{x\})} X^{-1}(\{y\}),$$

and these events are each single disjoint elements. Therefore,

$$\mathbb{P}(X^{-1}(h^{-1}(\{x\}))) = \sum_{y \in h^{-1}(\{x\})} \mathbb{P}(X^{-1}(\{y\})).$$

By the definition of  $\mathbb{P}_X$ ,

$$\mathbb{P}_X(\{y\}) = \mathbb{P}(X^{-1}(\{y\})).$$

Thus,

$$\mathbb{P}(Y^{-1}(\{x\})) = \sum_{y \in h^{-1}(\{x\})} \mathbb{P}_X(\{y\}) = \mathbb{P}_X(h^{-1}(\{x\})).$$

Finally, by definition of  $\tilde{\mathbb{P}}_1$ ,

$$\tilde{\mathbb{P}}_1(\{x\}) = \mathbb{P}_X(h^{-1}(\{x\})).$$

Thus,

$$\mathbb{P}_1(\{x\}) = \tilde{\mathbb{P}}_1(\{x\}),$$

and since this holds for every  $x \in E_1$ ,

$$\mathbb{P}_1 = \tilde{\mathbb{P}}_1.$$

□

## Question 2

Now, let  $X_1, X_2 : \Omega \rightarrow \mathbb{R}$  be discrete random variables with sets of values  $V_{X_1}, V_{X_2} \subset \mathbb{R}$ . Prove that for any function  $h : \mathbb{R}^2 \rightarrow \mathbb{R}$  we have

$$E(h(X_1, X_2)) = \sum_{(x_1, x_2) \in V_{X_1} \times V_{X_2}} h(x_1, x_2) f_{X_1, X_2}(x_1, x_2)$$

*Hint:* Use Question 1, taking  $E_0 = V_{X_1} \times V_{X_2}$ ,  $X = (X_1, X_2)$ , and  $h = h|_{V_{X_1} \times V_{X_2}} : V_{X_1} \times V_{X_2} \rightarrow V_Y$ .

Starting with the LHS, we have

$$\mathbb{E}(h(X_1, X_2)) = \sum_{y \in V_Y} y \mathbb{P}_1(\{h(X_1, X_2) = y\})$$

But now, we can use a key equality from the previous part. We know that

$$\mathbb{P}_1 = \tilde{\mathbb{P}}_1 \implies \mathbb{P}(Y^{-1}\{x\}) = \mathbb{P}_X(h^{-1}\{x\})$$

Hence, we can say that

$$\sum_{y \in V_Y} y \mathbb{P}(\{h(X_1, X_2) = y\}) = \sum_{y \in V_Y} y \sum_{(x_1, x_2) \in h^{-1}\{y\}} \mathbb{P}_X((X_1, X_2) = (x_1, x_2))$$

but from here, we can combine summands since we will cover every possible value in  $V_X \times V_Y$  and express  $y$  as  $h(x_1, x_2)$

$$\mathbb{E}(h(X_1, X_2)) = \sum_{(x_1, x_2) \in V_X \times V_Y} h(x_1, x_2) \mathbb{P}_X((X_1, X_2) = (x_1, x_2))$$

Hence,

$$E(h(X_1, X_2)) = \sum_{(x_1, x_2) \in V_{X_1} \times V_{X_2}} h(x_1, x_2) f_{X_1, X_2}(x_1, x_2)$$

□

# Conditional Distributions

## Question 3

Let  $X$  and  $Y$  be two discrete random variables on an abstract sample space  $\Omega$ , with possible values  $V_X = \{1, 2, 3\}$  and  $V_Y = \{1, 2\}$ , and with joint probability mass function

$$f_{X,Y}(x, y) = \frac{x + y}{21}.$$

Calculate the values of the conditional distribution functions

$$g(x | y) = P(X = x | Y = y) \quad \text{and} \quad h(y | x) = P(Y = y | X = x),$$

for  $x = 1, 2, 3$  and  $y = 1, 2$ .

We can first find the marginal probabilities for each  $X, Y$ , summing over all possible values of the opposite variable.

We can consider the  $\mathbb{P}(X = x)$  for each  $x \in V_X$ . For  $X = 1$ , we have

$$\begin{aligned} \mathbb{P}(X = 1) &= \sum_{y \in V_Y} f_{X,Y}(1, y) \\ &= \frac{1+1}{21} + \frac{1+2}{21} \\ &= \frac{2}{21} + \frac{3}{21} \\ &= \frac{5}{21} \end{aligned}$$

For  $X = 2$ , we have

$$\begin{aligned} \mathbb{P}(X = 2) &= \sum_{y \in V_Y} f_{X,Y}(2, y) \\ &= \frac{2+1}{21} + \frac{2+2}{21} \\ &= \frac{3}{21} + \frac{4}{21} \\ &= \frac{7}{21} \end{aligned}$$

For  $X = 3$ , we have

$$\begin{aligned} \mathbb{P}(X = 3) &= \sum_{y \in V_Y} f_{X,Y}(3, y) \\ &= \frac{3+1}{21} + \frac{3+2}{21} \\ &= \frac{4}{21} + \frac{5}{21} \\ &= \frac{9}{21} \end{aligned}$$

We can consider the  $\mathbb{P}(Y = y)$  for each  $y \in V_Y$ . For  $Y = 1$ , we have

$$\begin{aligned} \mathbb{P}(Y = 1) &= \sum_{x \in V_X} f_{X,Y}(x, 1) \\ &= \frac{1+1}{21} + \frac{2+1}{21} + \frac{3+1}{21} \\ &= \frac{2}{21} + \frac{3}{21} + \frac{4}{21} \\ &= \frac{9}{21} \end{aligned}$$

For  $Y = 2$ , we have

$$\begin{aligned} \mathbb{P}(Y = 2) &= \sum_{x \in V_X} f_{X,Y}(x, 2) \\ &= \frac{1+2}{21} + \frac{2+2}{21} + \frac{3+2}{21} \\ &= \frac{3}{21} + \frac{4}{21} + \frac{5}{21} \\ &= \frac{12}{21} \end{aligned}$$

Now, we can calculate the values of  $g(x | y)$  and  $h(y | x)$  using the fact that  $\mathbb{P}(X = x | Y = y) = \frac{\mathbb{P}(\{X=x\} \cap \{Y=y\})}{\mathbb{P}(Y=y)}$  and that in this formula, the numerator is simply our given function and the denominator is comprised of values we just calculated

**Calculating**  $g(x | y) = \mathbb{P}(X = x | Y = y)$ :  
For  $Y = 1$ :

$$\mathbb{P}(X = 1 | Y = 1) = \frac{f_{X,Y}(1,1)}{\mathbb{P}(Y = 1)} = \frac{\frac{2}{21}}{\frac{9}{21}} = \frac{2}{9},$$

$$\mathbb{P}(X = 2 | Y = 1) = \frac{f_{X,Y}(2,1)}{\mathbb{P}(Y = 1)} = \frac{\frac{3}{21}}{\frac{9}{21}} = \frac{3}{9} = \frac{1}{3},$$

$$\mathbb{P}(X = 3 | Y = 1) = \frac{f_{X,Y}(3,1)}{\mathbb{P}(Y = 1)} = \frac{\frac{4}{21}}{\frac{9}{21}} = \frac{4}{9}.$$

For  $Y = 2$ :

$$\mathbb{P}(X = 1 | Y = 2) = \frac{f_{X,Y}(1,2)}{\mathbb{P}(Y = 2)} = \frac{\frac{3}{21}}{\frac{12}{21}} = \frac{3}{12} = \frac{1}{4},$$

$$\mathbb{P}(X = 2 | Y = 2) = \frac{f_{X,Y}(2,2)}{\mathbb{P}(Y = 2)} = \frac{\frac{4}{21}}{\frac{12}{21}} = \frac{4}{12} = \frac{1}{3},$$

$$\mathbb{P}(X = 3 | Y = 2) = \frac{f_{X,Y}(3,2)}{\mathbb{P}(Y = 2)} = \frac{\frac{5}{21}}{\frac{12}{21}} = \frac{5}{12}.$$

**Calculating**  $h(y | x) = \mathbb{P}(Y = y | X = x)$ :  
For  $X = 1$ :

$$\mathbb{P}(Y = 1 | X = 1) = \frac{f_{X,Y}(1,1)}{\mathbb{P}(X = 1)} = \frac{\frac{2}{21}}{\frac{5}{21}} = \frac{2}{5},$$

$$\mathbb{P}(Y = 2 | X = 1) = \frac{f_{X,Y}(1,2)}{\mathbb{P}(X = 1)} = \frac{\frac{3}{21}}{\frac{5}{21}} = \frac{3}{5}.$$

For  $X = 2$ :

$$\mathbb{P}(Y = 1 | X = 2) = \frac{f_{X,Y}(2,1)}{\mathbb{P}(X = 2)} = \frac{\frac{3}{21}}{\frac{7}{21}} = \frac{3}{7},$$

$$\mathbb{P}(Y = 2 | X = 2) = \frac{f_{X,Y}(2,2)}{\mathbb{P}(X = 2)} = \frac{\frac{4}{21}}{\frac{7}{21}} = \frac{4}{7}.$$

For  $X = 3$ :

$$\mathbb{P}(Y = 1 | X = 3) = \frac{f_{X,Y}(3,1)}{\mathbb{P}(X = 3)} = \frac{\frac{4}{21}}{\frac{9}{21}} = \frac{4}{9},$$

$$\mathbb{P}(Y = 2 | X = 3) = \frac{f_{X,Y}(3,2)}{\mathbb{P}(X = 3)} = \frac{\frac{5}{21}}{\frac{9}{21}} = \frac{5}{9}.$$

Thus, we have calculated the values of each of those two functions, defining them on all necessary values in our set.

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## Question 4

Calculate the variances  $\sigma^2(X)$  and  $\sigma^2(Y)$ , the covariance

$$\sigma(X, Y) = E[(X - \mu_X)(Y - \mu_Y)],$$

and the correlation coefficient

$$\rho_{X,Y} = \frac{\sigma(X, Y)}{\sigma_X \sigma_Y}.$$

Let's first calculate both  $\mathbb{E}(X)$  and  $\mathbb{E}(Y)$ . This gives

$$\mu_X = \mathbb{E}(X) = \sum_{x \in V_X} x f_X(x) = \frac{5}{21} + \frac{14}{21} + \frac{27}{21} = \frac{46}{21}$$

and

$$\mu_Y = \mathbb{E}(Y) = \sum_{y \in V_Y} y f_Y(y) = \frac{9}{21} + \frac{24}{21} = \frac{33}{21}$$

Now, let's to find the variance, we also need  $\mathbb{E}(X^2), \mathbb{E}(Y^2)$

**For  $X$ :**

$$\mathbb{E}(X^2) = \sum_{x \in V_X} x^2 f_X(x)$$

**For  $Y$ :**

$$\mathbb{E}(Y^2) = \sum_{y \in V_Y} y^2 f_Y(y)$$

$$\begin{aligned}
&= \frac{5}{21} + \frac{28}{21} + \frac{81}{21} \\
&= \frac{114}{21}.
\end{aligned}$$

Thus, the variance of  $X$  is

$$\begin{aligned}
\sigma^2(X) &= \mathbb{E}(X^2) - \mu_X^2 \\
&= \frac{114}{21} - \left(\frac{46}{21}\right)^2 \\
&= \frac{2394}{441} - \frac{2116}{441} \\
&= \frac{278}{441}.
\end{aligned}$$

$$\begin{aligned}
&= \frac{9}{21} + \frac{48}{21} \\
&= \frac{57}{21}.
\end{aligned}$$

Thus, the variance of  $Y$  is

$$\begin{aligned}
\sigma^2(Y) &= \mathbb{E}(Y^2) - \mu_Y^2 \\
&= \frac{57}{21} - \left(\frac{33}{21}\right)^2 \\
&= \frac{1197}{441} - \frac{1089}{441} \\
&= \frac{108}{441} \\
&= \frac{12}{49}.
\end{aligned}$$

Now, to find the covariance, consider

$$\mathbb{E}[(X - \mu_X)(Y - \mu_Y)] = \mathbb{E}(XY) - \mu_X \mathbb{E}(Y) - \mu_Y \mathbb{E}(X) + \mu_X \mu_Y = \mathbb{E}(XY) - \mu_X \mu_Y.$$

This gives

$$\mathbb{E}(XY) = \sum_{x \in V_X} \sum_{y \in V_Y} xy f_{X,Y}(x, y).$$

Calculating term-by-term:

$$\begin{aligned}
x = 1, y = 1: & 1 \cdot 1 \cdot \frac{1+1}{21} = \frac{2}{21}, & x = 3, y = 1: & 3 \cdot 1 \cdot \frac{3+1}{21} = \frac{12}{21}, \\
x = 1, y = 2: & 1 \cdot 2 \cdot \frac{1+2}{21} = \frac{6}{21}, & x = 3, y = 2: & 3 \cdot 2 \cdot \frac{3+2}{21} = \frac{30}{21}, \\
x = 2, y = 1: & 2 \cdot 1 \cdot \frac{2+1}{21} = \frac{6}{21}, \\
x = 2, y = 2: & 2 \cdot 2 \cdot \frac{2+2}{21} = \frac{16}{21},
\end{aligned}$$

Thus, summing these terms we have:

$$\mathbb{E}(XY) = \frac{2 + 6 + 6 + 16 + 12 + 30}{21} = \frac{72}{21}.$$

Therefore, the covariance is

$$\sigma(X, Y) = \mathbb{E}(XY) - \mu_X \mu_Y = \frac{72}{21} - \frac{46 \cdot 33}{21^2} = \frac{72}{21} - \frac{1518}{441} = \frac{1512 - 1518}{441} = -\frac{6}{441} = -\frac{2}{147}.$$

Finally, the correlation coefficient is given by

$$\rho_{X,Y} = \frac{\sigma(X, Y)}{\sigma_X \sigma_Y} = \frac{-\frac{2}{147}}{\left(\sqrt{\frac{278}{441}}\right) \left(\sqrt{\frac{108}{441}}\right)} = \frac{-\frac{2}{147}}{\frac{2\sqrt{834}}{147}} = -\frac{1}{\sqrt{834}}.$$