Homework Assignment 1

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Problem 3.7. Suppose p(x, y, z), the joint probability mass function of the random variables X, Y, and Z, is given by

$$p(1,1,1) = \frac{1}{8}, \quad p(2,1,1) = \frac{1}{4},$$

$$p(1,1,2) = \frac{1}{8}, \quad p(2,1,2) = \frac{3}{16},$$

$$p(1,2,1) = \frac{1}{16}, \quad p(2,2,1) = 0,$$

$$p(1,2,2) = 0, \quad p(2,2,2) = \frac{1}{4}.$$

What is E[X|Y=2]? What is E[X|Y=2,Z=1]?

Solution. Recall that the conditional probability mass function of X given that Y = y is given by

$$p_{X|Y}(x|y) = P\{X = x|Y = y\} = \frac{P\{X = x, Y = y\}}{P\{Y = y\}}.$$

As a natural extension, we have that the conditional expectation of X given that Y = y is given by

$$E[X|Y = y] = \sum_{x} xP\{X = x|Y = y\} = \sum_{x} xp_{X|Y}(x|y).$$

Thus, in order to find the conditional expectation of X given that Y = 2, i.e. E[X|Y = 2], we first need to determine $p_{X|Y}(x|2)$. We note from the above joint probability mass function that

$$P\{Y=2\} = \sum_{x,z} p(x,2,z) = p(1,2,1) + p(2,2,1) + p(1,2,2) + p(2,2,2) = \frac{5}{16}.$$

Similarly, we have from the above joint probability mass function that

$$P{X = x, Y = 2} = \sum_{z} p(x, 2, z) = p(x, 2, 1) + p(x, 2, 2).$$

Thus, the conditional probability mass function of X given that Y=2 is given by

$$p_{X|Y}(x|2) = \frac{P\{X = x, Y = 2\}}{P\{Y = 2\}} = \begin{cases} \frac{p(1,2,1) + p(1,2,2)}{5/16} = \frac{1}{5} & \text{if } x = 1\\ \frac{p(1,2,1) + p(1,2,2)}{5/16} = \frac{4}{5} & \text{if } x = 2. \end{cases}$$

Using $p_{X|Y}(x|2)$, we readily see that

$$E[X|Y=2] = \sum_{x} x p_{X|Y}(x|2) = 1 \cdot p_{X|Y}(1|2) + 2 \cdot p_{X|Y}(2|2) = \frac{9}{5}.$$

In order to find the conditional expectation of X given that Y=2 and Z=1, i.e. E[X|Y=2,Z=1], we proceed in a similar manner as previously by first finding $p_{X|Y,Z}(x|2,1)$. We note from the above joint probability mass function that

$$P{Y = 2, Z = 1} = \sum_{x} p(x, 2, 1) = p(1, 2, 1) + p(2, 2, 1) = \frac{1}{16}$$

Similarly, we have from the above joint probability mass function that

$$P{X = x, Y = 2, Z = 1} = p(x, 2, 1).$$

Thus, the conditional probability mass function of X given that Y=2 and Z=1 is given by

$$p_{X|Y,Z}(x|2,1) = \frac{P\{X = x, Y = 2, Z = 1\}}{P\{Y = 2, Z = 1\}} = \begin{cases} \frac{p(1,2,1)}{1/16} = 1 & \text{if } x = 1\\ \frac{p(2,2,1)}{1/16} = 0 & \text{if } x = 2. \end{cases}$$

Using $p_{X|Y,Z}(x|2,1)$, we readily see that

$$E[X|Y=2,Z=1] = \sum_{x} x p_{X|Y,Z}(x|2,1) = 1 \cdot p_{X|Y,Z}(1|2,1) + 2 \cdot p_{X|Y,Z}(2|2,1) = 1.$$

Problem 3.8. An unbiased die is successively rolled. Let X and Y denote, respectively, the number of rolls necessary to obtain a six and a five. Find:

- a. E[X],
- b. E[X|Y=1],
- c. E[X|Y = 5].

Solution. The experiment of rolling a die, assuming the die is six-sided, has six possible outcomes: the die lands oriented such that the side with 1, 2, 3, 4, 5, or 6 pips is face-up. Assuming the die is unbiased, each outcome occurs with probability p = 1/6 and each trial of rolling the die is independent of any other trial. If X and Y denote, respectively, the number of rolls necessary to obtain a six and a five, then under the given assumptions, X and Y are both geometric random variables with parameter p = 1/6. The probability mass function for these random variables is given by $p(n) = (1-p)^{n-1}p = (5/6)^{n-1}(1/6)$.

a. Let Z be the random variable defined as Z = 1 if the result of the first roll is a six and Z = 0 if the result of the first roll is not a six. We may compute E[X] by conditioning on the variable Z. Note that, by conditioning, we obtain

$$\begin{split} E[X] &= \sum_z E[X|Z=z] P\{Z=z\} \\ &= \left[\frac{1}{6}\right] E[X|Z=1] + \left[\frac{5}{6}\right] E[X|Z=0]. \end{split}$$

If Z=1, then the result of the first roll is a six, so the number of rolls to obtain a six is clearly 1 and E[X|Z=1]=1. Likewise, if the result of the first roll is not a six, then the expected number of rolls to obtain a six given that the first roll is not a six is 1 more than the expected number of rolls to obtain a six so that E[X|Z=0]=1+E[X]. Therefore,

$$E[X] = \left[\frac{1}{6}\right] E[X|Z=1] + \left[\frac{5}{6}\right] E[X|Z=0]$$
$$= \frac{1}{6} + \left[\frac{1}{6}\right] (1 + E[X])$$

which implies that E[X] = 6.

b. We wish to find E[X|Y=1], i.e. the expected number of rolls to obtain a six given that the first roll is a five. Using the same reasoning as in part a, we know that the expected number of rolls to obtain a six given that the first roll is not a six (it's a five) is 1 more than the expected number of rolls to obtain a six. Therefore,

$$E[X|Y = 1] = 1 + E[X] = 7$$

where we used the result previously obtained that E[X] = 6.

c. In order to calculate E[X|Y=y] for some y>1, we first compute $p_{X|Y}(x|y)$. Suppose that Y=y for some y>1. From this we gather that the first y-1 trials result in not rolling a five while the y-th trial results in rolling a five.

As a consequence, if X = x where x < y then the first x trials have only five possible outcomes with the x-th trial resulting in a success out of those five outcomes. Thus,

$$P\{X = x | Y = y\} = \frac{1}{5} \left[\frac{4}{5}\right]^{x-1},$$

i.e. for x < y the conditional probability that X = x given that Y = y is the probability mass function of a geometric random variable with parameter p = 1/5.

Note that if X = x where x = y, then

$$P\{X = x | Y = y\} = 0$$

since it cannot happen that on the y-th trial the outcome of the trial is that both a five and a six were rolled.

Finally, if X = x where x > y, then as mentioned, the first y - 1 trials do not result in a five, but after the y-th trial the result obtained can in fact be a five. Thus, the first y - 1 failures each occur with probability 4/5 while the y-th failure occurs with probability 1. However, after that, the failures of the trials y + 1 through x - 1 all occur with probability 5/6 since it is possible for the die to roll a five during these trials. On the x-th trial the trial succeeds with probability 1/6. Thus, if x > y, then

$$P\{X = x | Y = y\} = \frac{1}{6} \left[\frac{4}{5}\right]^{y-1} \left[\frac{5}{6}\right]^{x-y-1}.$$

Combining the above statements, we see that the conditional probability mass function that X = x given that Y = y with y > 1 is given by

$$p_{X|Y}(x|y) = \begin{cases} \frac{1}{5} \left[\frac{4}{5} \right]^{x-1} & \text{if } x < y \\ 0 & \text{if } x = y \\ \frac{1}{6} \left[\frac{4}{5} \right]^{y-1} \left[\frac{5}{6} \right]^{x-y-1} & \text{if } x > y \end{cases}$$

Therefore, we have that the expected value of X given that Y=5 is

$$E[X|Y=5] = \sum_{x=1}^{\infty} x p_{X|Y}(x|5)$$

$$= \frac{1}{5} \sum_{x=1}^{4} x \left[\frac{4}{5} \right]^{x-1} + \frac{1}{6} \left[\frac{4}{5} \right]^{4} \sum_{x=6}^{\infty} x \left[\frac{5}{6} \right]^{x-6}$$

$$= \frac{3637}{625} \approx 5.82.$$

Problem 3.9. Show in the discrete case that if X and Y are independent, then

$$E[X|Y=y] = E[X]$$
 for all y.

Solution. Suppose that X and Y are discrete, independent random variables. Then, due to the independence of the random variables, we know that

$$P\{X = x | Y = y\} = \frac{P\{X = x, Y = y\}}{P\{Y = y\}}$$

$$= \frac{P\{X = x\}P\{Y = y\}}{P\{Y = y\}}$$

$$= P\{X = x\}. \tag{1}$$

Therefore, combining the definition of conditional expectation for discrete random variables and result (1), we have that for any y,

$$E[X|Y = y] = \sum_{x} xP\{X = x|Y = y\} = \sum_{x} xP\{X = x\} = E[X]$$

and we are done. \Box

Problem 3.10. Suppose X and Y are independent continuous random variables. Show that

$$E[X|Y=y] = E[X]$$
 for all y.

Solution. Suppose that X and Y are continuous, independent random variables with probability density functions $f_X(x)$ and $f_Y(y)$, respectively. The conditional probability density function of X given that Y = y is given by

$$f_{X|Y}(x|y) = \frac{f(x,y)}{f_Y(y)}$$

where f(x, y) is the joint probability density function of X and Y. Due to the independence of the random variables X and Y, we have that $f(x, y) = f_X(x)f_Y(y)$. Thus, if X and Y are independent, then the conditional probability density function of X given that Y = y is

$$f_{X|Y}(x|y) = \frac{f(x,y)}{f_Y(y)} = \frac{f_X(x)f_Y(y)}{f_Y(y)} = f_X(x).$$
 (2)

Therefore, combining the definition of conditional expectation for continuous random variables and result (2), we have that for any y,

$$E[X|Y=y] = \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx = \int_{-\infty}^{\infty} x f_X(x) dx = E[X]$$

and we are done. \Box

Problem 3.13. Let X be exponential with mean $1/\lambda$; that is,

$$f_X(x) = \lambda e^{-\lambda x}, \quad 0 < x < \infty.$$

Find E[X|X > 1].

Solution. Suppose that X is an exponential random variable with mean $1/\lambda$. Let Y be the discrete random variable defined as

$$Y = \begin{cases} 1 & \text{if } X > 1 \\ 0 & \text{if } 0 < X \le 1 \end{cases}$$

with probability mass function

$$p_Y(y) = \begin{cases} P\{X > 1\} & \text{if } Y = 1\\ P\{X \le 1\} & \text{if } Y = 0 \end{cases}$$

From these definitions of X and Y, we see that the conditional density function of X given Y = 1 is given by

$$f_{X|Y}(x|1) = \frac{f(x,1)}{p_Y(1)} = \begin{cases} \frac{f_X(x)}{P\{Y=1\}} & \text{if } x > 1\\ 0 & \text{if } 0 < x \le 1 \end{cases}$$

where we know that if x > q then

$$\frac{f_X(x)}{P\{Y=1\}} = \frac{\lambda e^{-\lambda x}}{1 - P\{X \le 1\}}$$

$$= \frac{\lambda e^{-\lambda x}}{P\{X > 1\}}$$

$$= \frac{\lambda e^{-\lambda x}}{1 - \int_0^1 \lambda e^{-\lambda x} dx}$$

$$= \frac{\lambda e^{-\lambda x}}{e^{-\lambda}}.$$

Thus, by definition, we now have that

$$\begin{split} E[X|X>1] &= E[X|Y=1] \\ &= \int_{-\infty}^{\infty} x f_{X|Y}(x|1) dx \\ &= \int_{1}^{\infty} x \frac{\lambda e^{-\lambda x}}{e^{-\lambda}} dx \\ &= \lambda e^{\lambda} \int_{1}^{\infty} x e^{-\lambda x} dx \\ &= \lambda e^{\lambda} \left[\frac{e^{-\lambda}(1+\lambda)}{\lambda^2} \right] \\ &= \frac{1+\lambda}{\lambda}. \end{split}$$

Problem 3.14. Let X be uniform over (0,1). Find E[X|X<1/2]. Solution. \Box