

Please provide complete and well-written solutions to the following exercises.

Due May 24, in the discussion section.

## Homework 7

**Exercise 1.** Let  $X, Y$  and  $Z$  be independent geometric random variables with the same parameter  $0 < p < 1$ . Let  $k, n$  be nonnegative integers. Compute  $\mathbf{P}(X = k | X + Y + Z = n)$ .

**Exercise 2.** Let  $X, Y, Z$  be discrete random variables. Let  $A \subseteq \Omega$ . Let  $c \in \mathbf{R}$ . Show that conditional expectation is linear; that is, show that  $\mathbf{E}(X + Z | A) = \mathbf{E}(X | A) + \mathbf{E}(Z | A)$  and  $\mathbf{E}(cX | A) = c\mathbf{E}(X | A)$ . (Recall that Expectation itself is linear as well.) (Hint: it may be useful to write  $\mathbf{P}(\{Z + X = t\} \cap A) = \sum_{z, x \in \mathbf{R}: z+x=t} \mathbf{P}(\{Z = z\} \cap \{X = x\} \cap A)$ .)

Now, let  $f(y) = \mathbf{E}(X | Y = y)$  for any  $y \in \mathbf{R}$ . Then  $f: \mathbf{R} \rightarrow \mathbf{R}$  is a function. In more advanced probability classes, we consider the random variable  $f(Y)$ , which is denoted by  $\mathbf{E}(X | Y)$ . Show that  $\mathbf{E}(X + Z | Y) = \mathbf{E}(X | Y) + \mathbf{E}(Z | Y)$ .

Then, show that  $\mathbf{E}[\mathbf{E}(X | Y)] = \mathbf{E}(X)$ . (That is, you should show that  $\mathbf{E}f(Y) = \mathbf{E}(X)$ .) So, understanding  $\mathbf{E}(X | Y)$  can help us to compute  $\mathbf{E}(X)$ .

**Exercise 3.** Give an example of two random variables  $X, Y$  that are independent. Prove that these random variables are independent.

Give an example of two random variables  $X, Y$  that are not independent. Prove that these random variables are not independent.

Finally, find two random variables  $X, Y$  such that  $\mathbf{E}(XY) \neq \mathbf{E}(X)\mathbf{E}(Y)$ .

**Exercise 4.** Is it possible to have a random variable  $X$  such that  $X$  is independent of  $X$ ? Either find such a random variable  $X$ , or prove that it is impossible to find such a random variable  $X$ .

**Exercise 5.** Let  $0 < p < 1$ . Let  $n$  be a positive integer. Let  $X_1, \dots, X_n$  be pairwise independent Bernoulli random variables. Compute the expected value of

$$S_n = \frac{X_1 + \dots + X_n}{n}.$$

Then, compute the variance of  $S_n - \mathbf{E}(S_n)$ . Describe in words what this variance computation tells you as  $n \rightarrow \infty$ . Particularly, what does  $S_n$  “look like” as  $n \rightarrow \infty$ ? (Think about what we found in the Example, “Playing Monopoly Forever”. Also, consider the following statistical interpretation. Suppose each  $X_i$  is the result of some poll of person  $i$ , where  $i \in \{1, \dots, n\}$ . Suppose that each person’s response is a Bernoulli random variable with parameter  $p$ , and each person’s response is independent of each other person’s response. Then  $S_n$  is the average of the results of the poll. If  $S_n - \mathbf{E}(S_n)$  has small variance, then our poll is very accurate. So, how accurate is the poll as  $n \rightarrow \infty$ ? Note that the accuracy of the poll does *not* depend

on the size of the population you are sampling from!) (Here we are thinking of choosing  $n$  random people from the population, so that  $n$  is the number of people that participate in the poll, whereas the population could have many more than  $n$  people.)

**Exercise 6.** Let  $X$  and  $Y$  be discrete random variables on a sample space  $\Omega$ , and let  $\mathbf{P}$  be a probability law on  $\Omega$ . Assume that  $X$  and  $Y$  are independent. Assume that  $X$  and  $Y$  take a finite number of values. Let  $f, g: \mathbf{R} \rightarrow \mathbf{R}$  be functions. Show that

$$\mathbf{E}(f(X)g(Y)) = \mathbf{E}(f(X))\mathbf{E}(g(Y)).$$

**Exercise 7.** Find three random variables  $X_1, X_2, X_3$  such that:  $X_1$  and  $X_2$  are independent;  $X_1$  and  $X_3$  are independent;  $X_2$  and  $X_3$  are independent; but such that  $X_1, X_2, X_3$  are not independent.

**Exercise 8.** Let  $0 < p < 1$ . Let  $X_1, \dots, X_n$  be independent Bernoulli random variables with parameter  $p$ . Let  $S_n = \sum_{i=1}^n X_i$ . A moment generating function can help us to compute moments in various ways. Fix  $t \in \mathbf{R}$  and compute the moment generating function of  $X_i$  for each  $i \in \{1, \dots, n\}$ . That is, show that

$$\mathbf{E}e^{tX_i} = (1 - p) + pe^t.$$

Then, using the product formula for independent random variables, show that

$$\mathbf{E}e^{tS_n} = [(1 - p) + pe^t]^n.$$

By differentiating the last equality at  $t = 0$ , and using the power series expansion of the exponential function, compute  $\mathbf{E}S_n$  and  $\mathbf{E}S_n^2$ .

**Exercise 9.**  $X_1, \dots, X_n$  be independent discrete random variables. Show that

$$\mathbf{P}(X_1 \leq x_1, \dots, X_n \leq x_n) = \prod_{i=1}^n \mathbf{P}(X_i \leq x_i), \quad \forall x_1, \dots, x_n \in \mathbf{R}.$$

**Exercise 10.** Verify that  $\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = 1$ . (Hint: let  $T = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$ . It suffices to show that  $T^2 = 1$ , since  $T > 0$ .)