

EE 131A  
Probability and Statistics  
Instructor: Lara Dolecek

Homework 2 Solution  
Monday, January 11, 2021  
Due: Wednesday, January 20, 2021  
before class begins  
levtauz@ucla.edu  
debarnabucla@ucla.edu

**Please upload your homework to Gradescope by January 20, 3:59 pm.**  
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**You may type your homework or scan your handwritten version. Make sure all the work is discernible.**

Reading: Chapters 2.4-2.5 & 3 of *Probability, Statistics, and Random Processes* by A. Leon-Garcia

1. Roll two fair dice independently. In terms of the possible outcomes, define the events:

$$A = \{\text{First die is 1, 2 or 3}\}$$

$$B = \{\text{First die is 2, 3 or 6}\}$$

$$C = \{\text{Sum of outcomes is 9}\}$$

Are  $A$ ,  $B$ , and  $C$  mutually independent? Hint: Three events  $A$ ,  $B$ , and  $C$  are independent if all the four following constraints hold:

$$P(A \cap B) = P(A)P(B),$$

$$P(A \cap C) = P(A)P(C),$$

$$P(B \cap C) = P(B)P(C),$$

$$P(A \cap B \cap C) = P(A)P(B)P(C).$$

**Solution:** Sample space is  $S = \{(i, j) | 1 \leq i, j \leq 6\}$ . And,

$$A = \{(i, j) : i = 1, 2, 3 \text{ and } 1 \leq j \leq 6\},$$

$$B = \{(i, j) : i = 2, 3, 6 \text{ and } 1 \leq j \leq 6\},$$

$$C = \{(i, j) : i + j = 9\} = \{(3, 6), (4, 5), (5, 4), (6, 3)\}.$$

So,  $|S| = 36$ ,  $|A| = 18$ ,  $|B| = 18$ , and  $|C| = 4$ . Since the two fair dice rolls are independent,  $P(A) = \frac{18}{36} = \frac{1}{2}$ ,  $P(B) = \frac{18}{36} = \frac{1}{2}$ , and  $P(C) = \frac{4}{36} = \frac{1}{9}$ .

$$A \cap B = \{(i, j) | i = 2, 3 \text{ and } 1 \leq j \leq 6\},$$

$$A \cap C = \{(3, 6)\},$$

$$B \cap C = \{(3, 6), (6, 3)\},$$

$$A \cap B \cap C = \{(3, 6)\}.$$

Hence,  $P(A \cap B) = \frac{12}{36} = \frac{1}{3}$ ,  $P(A \cap C) = \frac{1}{36}$ ,  $P(B \cap C) = \frac{2}{36} = \frac{1}{18}$ ,  $P(A \cap B \cap C) = \frac{1}{36}$ . Three events  $A$ ,  $B$ , and  $C$  are independent if all the four following constraints hold:

$$P(A \cap B) = P(A)P(B), \quad (1)$$

$$P(A \cap C) = P(A)P(C), \quad (2)$$

$$P(B \cap C) = P(B)P(C), \quad (3)$$

$$P(A \cap B \cap C) = P(A)P(B)P(C). \quad (4)$$

Observe that (1) and (2) do not hold. Therefore,  $A$ ,  $B$ , and  $C$  are **not** independent.

2. Assume there are 5 jars numbered 1 to 5. The  $i^{\text{th}}$  jar contains  $i$  black balls,  $6 - i$  red balls, and 5 green balls. A jar is selected uniformly at random and a ball is selected uniformly at random from that jar. Let the events  $B$ ,  $R$ , and  $G$  represent the events that a black, red, or green ball is chosen, respectively. Let  $J_k$  represent the event that the  $k^{\text{th}}$  jar is chosen.

- (a) What is  $P(B|J_k)$ ?

**Solution:** By the problem definition,  $P(B|J_k) = \frac{k}{11}$ .

- (b) What is  $P(G)$ ,  $P(B)$ , and  $P(R)$ ?

**Solution:** Regardless of which jar is chosen, the green balls always make up 5 of the 11 available balls. Hence,  $P(G) = \frac{5}{11}$ .

By symmetry,  $P(B) = P(R)$ . Therefore,

$$\begin{aligned} 1 &= P(B) + P(R) + P(G) = 2 \cdot P(B) + \frac{5}{11} \\ \implies P(B) &= \frac{3}{11} \end{aligned}$$

Hence,  $P(B) = P(R) = \frac{3}{11}$ .

- (c) Given that the selected ball is black, what is the probability that the ball came from the  $k$ th jar, i.e.  $P(J_k|B)$ ?

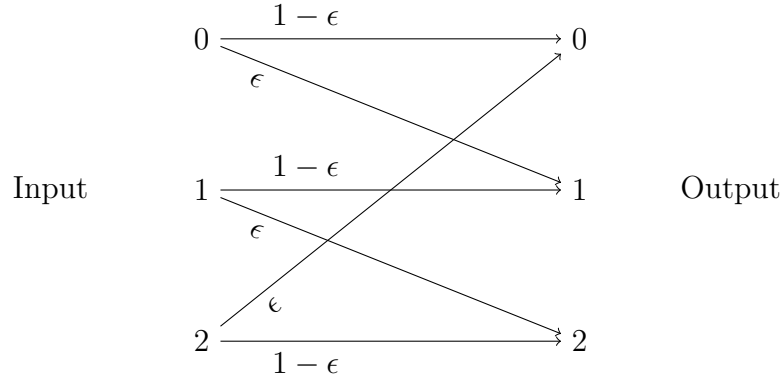
**Solution:** By Bayes rule, we have

$$P(J_k|B) = \frac{P(B|J_k)P(J_k)}{P(B)}$$

By using the values determined in previous parts, we get

$$P(J_k|B) = \frac{\frac{k}{11} \cdot \frac{1}{5}}{\frac{3}{11}} = \frac{k}{15}$$

3. A ternary communication channel is shown in the figure. Assume that input symbols 0, 1, and 2 are chosen for transmission with probabilities  $\frac{1}{4}$ ,  $\frac{1}{2}$ , and  $\frac{1}{4}$ , respectively.



- (a) Calculate the probability of each output.

**Solution:**

$$\begin{aligned}
 P(\text{Output} = 0) &= P(\text{Output} = 0 | \text{Input} = 0)P(\text{Input} = 0) \\
 &\quad + P(\text{Output} = 0 | \text{Input} = 1)P(\text{Input} = 1) \\
 &\quad + P(\text{Output} = 0 | \text{Input} = 2)P(\text{Input} = 2) \\
 &= \frac{1}{4}(1 - \epsilon) + 0 \cdot \frac{1}{2} + \frac{1}{4}\epsilon = \frac{1}{4}
 \end{aligned}$$

$$\begin{aligned}
 P(\text{Output} = 1) &= P(\text{Output} = 1 | \text{Input} = 0)P(\text{Input} = 0) \\
 &\quad + P(\text{Output} = 1 | \text{Input} = 1)P(\text{Input} = 1) \\
 &\quad + P(\text{Output} = 1 | \text{Input} = 2)P(\text{Input} = 2) \\
 &= \frac{1}{4}\epsilon + \frac{1}{2}(1 - \epsilon) + 0 \cdot \frac{1}{4} = \frac{2 - \epsilon}{4}
 \end{aligned}$$

$$\begin{aligned}
 P(\text{Output} = 2) &= P(\text{Output} = 2 | \text{Input} = 0)P(\text{Input} = 0) \\
 &\quad + P(\text{Output} = 2 | \text{Input} = 1)P(\text{Input} = 1) \\
 &\quad + P(\text{Output} = 2 | \text{Input} = 2)P(\text{Input} = 2) \\
 &= 0 \cdot \frac{1}{4} + \frac{1}{2}\epsilon + \frac{1}{4}(1 - \epsilon) = \frac{1}{4} + \frac{1}{4}\epsilon = \frac{1 + \epsilon}{4}
 \end{aligned}$$

- (b) Given that the output was 1, what is the probability that the input was 0? 1? 2?

**Solution:**

We want to figure out  $P(\text{Input} = k | \text{Output} = 1)$ . By Bayes rule, we get

$$P(\text{Input} = k | \text{Output} = 1) = \frac{P(\text{Output} = 1 | \text{Input} = k)P(\text{Input} = k)}{P(\text{Output} = 1)}.$$

All the necessary terms were calculated in part (a) which gives us the result

$$\begin{aligned}
 P(\text{Input} = 0 | \text{Output} = 1) &= \frac{\frac{1}{4}\epsilon}{\frac{2-\epsilon}{4}} = \frac{\epsilon}{2-\epsilon} \\
 P(\text{Input} = 1 | \text{Output} = 1) &= \frac{\frac{1}{2}(1-\epsilon)}{\frac{2-\epsilon}{4}} = \frac{2-2\epsilon}{2-\epsilon} \\
 P(\text{Input} = 2 | \text{Output} = 1) &= 0
 \end{aligned}$$

4. A family has 5 natural children and has adopted 2 girls. Each natural child has equal probability of being a girl or a boy, independent of the other children. Find the PMF of the number of girls out of the 7 children.

**Solution:** First, let  $N$  be the number of natural children that are girls. Let us first consider how we would get the pmf of  $N$ . Note that  $N$  can only take values in  $\{0, 1, 2, 3, 4, 5\}$  since there are only 5 natural children. Now, consider the probability that  $N = k$ . This means that we want to figure out the probability that  $k$  out of the 5 children were born as girls and that  $5-k$  were born as boys. There are  $\binom{5}{k}$  ways to select  $k$  out of 5 children to be girls and that the probability that those  $k$  children are girls and that the other  $5-k$  children are boys is  $(P(\text{Child is girl}))^k (P(\text{Child is boy}))^{5-k} = \frac{1}{2^k} \frac{1}{2^{5-k}} = \frac{1}{2^5}$ .

Combining this together, we get the pmf of  $N$  as

$$P(N = k) = \begin{cases} \binom{5}{k} \frac{1}{2^5} & , \text{if } 0 \leq k \leq 5 \\ 0 & \text{else} \end{cases}$$

This distribution is known as a binomial distribution.

Since there are already 2 definite girls, we just need to shift this distribution. Thus, the pmf for  $G$  which is the number of girls out of the 7 children is

$$P(G = g) = \begin{cases} \binom{5}{g-2} \frac{1}{2^5} & , \text{if } 2 \leq g \leq 7 \\ 0 & \text{else} \end{cases}$$

5. Throw a pair of six-sided dice. Let  $X_1$  be the number of dots on the resulting face of the first die and let  $X_2$  be the number of dots on the resulting face of the second die. Let  $Z = X_1 + X_2$  be the sum of the two dice rolls.

(a) What is the pmf of  $Z$ ?

**Solution:**

$$P(Z = z) = \begin{cases} \frac{z-1}{36} & z \in \{2, 3, 4, 5, 6, 7\} \\ \frac{13-z}{36} & z \in \{8, 9, 10, 11, 12\} \\ 0 & \text{otherwise} \end{cases}$$

(b) What is  $\mathbb{E}[Z]$ ?  $Var(Z)$ ? **Solution:** By linearity of expectation, we get

$$\begin{aligned}\mathbb{E}[Z] &= \mathbb{E}[X_1 + X_2] \\ &= \mathbb{E}[X_1] + \mathbb{E}[X_2] \\ &= 3.5 + 3.5 = 7\end{aligned}$$

Note that  $X_1$  and  $X_2$  are independent, thus we get

$$\begin{aligned}Var(Z) &= Var(X_1 + X_2) \\ &= Var(X_1) + Var(X_2) \\ &= \frac{35}{12} + \frac{35}{12} = \frac{70}{12}\end{aligned}$$

(c) Given that  $Z = 10$ , what is the probability that  $X_1 = k$  for  $k \in \{1, 2, 3, 4, 5, 6\}$ ?

**Solution:**

By Bayes rule,

$$P(X_1 = k|Z = 10) = \frac{P(Z = 10|X_1 = k)P(X_1 = k)}{P(Z = 10)}.$$

From part (a), we know  $P(Z = 10) = \frac{3}{36}$ . Additionally, we know  $P(X_1 = k) = \frac{1}{6}$ . Due to  $Z = X_1 + X_2$  and the independence of  $X_1$  and  $X_2$ , we can write  $P(Z = 10|X_1 = k) = P(X_2 = 10 - X_1|X_1 = k) = P(X_2 = 10 - k)$ . Note that

$$P(X_2 = 10 - k) = \begin{cases} \frac{1}{6} & k \in \{4, 5, 6, 7, 8, 9\} \\ 0 & \text{otherwise} \end{cases}.$$

Combining these results together, we get

$$P(X_1 = k|Z = 10) = \begin{cases} \frac{1}{3} & k \in \{4, 5, 6\} \\ 0 & \text{otherwise} \end{cases}$$

6. Assume that we flip a biased coin with probability of heads being  $p$  until a 2nd head is seen. Let  $X$  be the number of flips up until and including the flip that has the 2nd head. What is the pmf and expectation of  $X$ ?

**Solution:**

Suppose  $X = n$ . Then the  $n^{th}$  flip must be a head. Of the first  $(n - 1)$  flips, there must be exactly 1 head. Since there are  $(n - 1)$  ways of getting exactly 1 head in  $(n - 1)$  flips, the probability that  $X = n$  is

$$P(X = n) = \begin{cases} (n - 1)p^2(1 - p)^{n-2} & , n = 2, 3, 4, 5, \dots \\ 0 & else \end{cases}$$

Now, we find the expectation of  $X$ .

$$\begin{aligned}
\mathbb{E}[X] &= \sum_{n=2}^{\infty} n \cdot P(X = n) \\
&= \sum_{n=2}^{\infty} n(n-1)p^2(1-p)^{n-2} \\
&= p^2 \sum_{n=2}^{\infty} n(n-1)(1-p)^{n-2} \\
&= p^2 \frac{d^2}{dp^2} \left( \sum_{n=2}^{\infty} (1-p)^n \right) && \text{Differentiation Trick} \\
&= p^2 \frac{d^2}{dp^2} \left( \frac{(1-p)^2}{1-(1-p)} \right) && \text{Geometric Sum Formula} \\
&= p^2 \frac{d^2}{dp^2} \left( \frac{1-2p+p^2}{p} \right) \\
&= p^2 \frac{d^2}{dp^2} \left( \frac{1}{p} - 2 + p \right) \\
&= p^2 \left( \frac{2}{p^3} \right) = \frac{2}{p}
\end{aligned}$$

Another way we could have arrived at this expectation is by using the linearity of expectation. Observe that the number of flips until the first head arrives is a geometric random variable regardless of when you start counting. As such, the number of flips from the first head to the second head is also a geometric random variable. Hence,  $X$  can be written as  $X = Y_1 + Y_2$  where  $Y_1$  and  $Y_2$  are geometric random variables with parameter  $p$ .

Hence,

$$\mathbb{E}[X] = \mathbb{E}[Y_1] + \mathbb{E}[Y_2] = \frac{1}{p} + \frac{1}{p} = \frac{2}{p}$$