

## C.1 Discrete Random Variables

**PROBLEM 1.** The  $\text{Im } Z$  is discrete because  $\text{Im } X$  and  $\text{Im } Y$  are discrete sets.

Let  $z \in \mathbb{R}$ . If  $\{Z = z\} = \emptyset$ , then  $\{Z = z\}$  is an event because the set  $\emptyset$  is always an event. Assume that  $\{Z = z\} \neq \emptyset$ . We have to consider two cases.

i)  $z = 0$ . In this case, the only way that  $Z(s) = 0$  is if  $X(s) = 0$  or  $Y(s) = 0$ . Therefore,

$$\{Z = 0\} = \{X = 0\} \cup \{Y = 0\}.$$

Since  $\{X = 0\}$  and  $\{Y = 0\}$  are events, we conclude that  $\{Z = 0\}$  are events (recall that, by assumption,  $X$  and  $Y$  are discrete random variables).

ii)  $z \neq 0$ . In this case, the functions  $X$  and  $Y$  can't take the value 0. If  $s \in \{Z = z\}$ , then  $X(s)Y(s) = Z(s) = z$ . Therefore  $X(s) = z/Y(s)$ . Let  $y = Y(s)$ . Then  $X(s) = z/y$  and  $Y(s) = y$ . In other words,  $s \in \{X = z/y\} \cap \{Y = y\}$ . On the other hand, if  $s \in \{X = z/y\} \cap \{Y = y\}$ , then  $X(s) = z/y$  and  $Y(s) = y$ . Therefore,  $Z(s) = X(s)Y(s) = (z/y)y = z$  and then  $s \in \{Z = z\}$ . In summary, we have just proved that

$$\{Z = z\} = \bigcup_{y \in \text{Im } Y, y \neq 0} \left( \{X = z/y\} \cap \{Y = y\} \right).$$

For a given  $z \in \mathbb{R}$ ,  $z \neq 0$  and  $y \in \text{Im } Y$ , the event  $\{X = z/y\} \cap \{Y = y\}$  is an event because  $X$  and  $Y$  are discrete random variables and  $\mathcal{A}$  is an event space. Thus, a countable union of these events will remain an event. Hence,  $\{Z = z\}$  is an event.

In each case,  $\{Z = z\}$  is an event. The map  $Z$  satisfies condition (a) and (b) in Definition C.1 and therefore  $Z$  is a discrete random variable.  $\triangle$

**PROBLEM 2.** We have  $\text{Im}(1_A) = \{0, 1\}$ , which is a finite set (therefore discrete).

Let  $x \in \mathbb{R}$ . We have three cases to consider.

1.  $x = 0$ . In this case,  $\{1_A = 0\} = \overline{A}$ . Since  $A$  is an event, we know that  $\overline{A}$  is also an event. Therefore,  $\{1_A = 0\}$  is an event.
2.  $x = 1$ . In this case,  $\{1_A = 1\} = A$  and  $A$  is an event. Hence,  $\{1_A = 1\}$  is an event.
3.  $x \neq 0$  and  $x \neq 1$ . In this case,  $\{1_A = x\} = \emptyset$  because there is no  $s$  such that  $1_A(s) = x$  (the only possible values are 0 and 1 for  $1_A$ ). Since  $\emptyset$  is an event,  $\{1_A = x\}$  is an event.  $\triangle$

**PROBLEM 3.**

- a) Let  $x \in \mathbb{R}$ . If  $\{X \leq x\} = \emptyset$ , then  $\{X \leq x\}$  is an event. Assume that  $\{X \leq x\} \neq \emptyset$ . Since  $X$  is a discrete random variable, the set  $\text{Im } X$  is discrete. This means there are only a countable values of  $\text{Im } X$  that can be smaller than the number  $x$ . List them in decreasing

order, say  $x_1, x_2, x_3, \dots$ , with  $x_i \geq x_j$ , when  $i \leq j$  and  $x_j \leq x$  for any  $j$ . Therefore, we can write

$$\{X \leq x\} = \bigcup_{j=1}^{\infty} \{X = x_j\}.$$

The map  $X$  is a discrete random variable. Therefore, each set  $\{X = x_j\}$  is an event and this implies that  $\bigcup_{j=1}^{\infty} \{X = x_j\}$  is an event. Hence,  $\{X \leq x\}$  is an event.

b) Let  $x \in \mathbb{R}$ . We can write

$$\{X < x\} = \{X \leq x\} \cap \overline{\{X = x\}}.$$

In other words, the set  $\{X < x\}$  is the set of  $s \in S$  that belong to  $\{X \leq x\}$  but are not in  $\{X = x\}$ . From part a), the set  $\{X \leq x\}$  is an event and from the fact that  $X$  is assumed to be a discrete random variable,  $\{X = x\}$  is an event. Therefore,  $\{X \leq x\} \cap \overline{\{X = x\}}$  is an event and hence  $\{X < x\}$  is an event.

c) We have

$$\{X \geq x\} = \overline{\{X < x\}}.$$

From part b), we know that  $\{X < x\}$  is an event, hence  $\{X \geq x\}$  is also an event.

d) We have

$$\{X > x\} = \overline{\{X \leq x\}}.$$

From part c), we know that  $\{X \leq x\}$  is an event, hence  $\{X > x\}$  is also an event.

**PROBLEM 4.** Assume that  $X$  is a discrete random variable. Then  $\text{Im } X$  is discrete and  $\{X = x\}$  is an event for every  $x \in \mathbb{R}$ . From Problem 3, part a), the set  $\{X \leq x\}$  is an event. Therefore, conditions a) and b) in the statement are satisfied.

Assume that the two conditions in the statement are satisfied. Then, in particular,  $\text{Im } X$  is discrete. Also, for an  $x \in \mathbb{R}$ , the set  $\{X > x\}$  is an event because it is the complement of the event  $\{X \leq x\}$ . Also, for an  $x \in \mathbb{R}$ , we have

$$\{X < x\} = \bigcup_{j=1}^{\infty} \left\{X \leq x - \frac{1}{j}\right\}.$$

This is a countable unions of the events  $\{X \leq x - \frac{1}{j}\}$  and therefore  $\{X < x\}$  is an event. But also  $\{X \geq x\}$  is also an event because it is the complement of  $\{X < x\}$ . Let  $x \in \mathbb{R}$ . We can write

$$\{X = x\} = \{X \leq x\} \cap \{X \geq x\},$$

the intersection of two events! So  $\{X = x\}$  is also an event. Hence  $X$  is a discrete random variable.

## C.2 Probability Mass Functions

**PROBLEM 5.** The probability measure  $P$  on  $S$  is given by

$$P(\{r\}) = \frac{2}{5}, \quad P(\{b\}) = \frac{2}{5}, \quad P(\{y\}) = \frac{1}{5}.$$

The function  $X : S \rightarrow \mathbb{R}$  is given by  $X(\{r\}) = -10$ ,  $X(\{b\}) = 10$ , and  $X(\{y\}) = 20$ . Therefore, we have

- $p_X(-10) = P(X = -10) = P(\{r\}) = \frac{2}{5}$ .
- $p_X(10) = P(X = 10) = P(\{b\}) = \frac{2}{5}$ .
- $p_X(20) = P(X = 20) = P(\{y\}) = \frac{1}{5}$ .
- $p_X(x) = 0$ , for  $x \neq -10, 10, 20$ .

**PROBLEM 6.** A child may or may not identify properly the picture. Let  $w_1, w_2, w_3$  be the words corresponding to the animal in picture  $p_1, p_2, p_3$ . A child will identify correctly a picture if the word  $w_i$  is put under the picture  $p_i$ . Therefore, we can identify an outcome as an ordered list of three symbols from  $\{w_1, w_2, w_3\}$ . For example,  $w_1w_2w_3$  means that the child identified the animal in picture  $p_1$  as  $w_1$ , in picture  $p_2$  as  $w_2$ , and in picture  $p_3$  as  $w_3$ . Therefore, the sample space is

$$S = \{w_1w_2w_3, w_1w_3w_2, w_2w_1w_3, w_3w_2w_1, w_3w_1w_2, w_2w_1w_3\}.$$

If we just keep the numbers

$$S = \{123, 132, 213, 321, 312, 231\}.$$

Each outcome are equally likely to happen, so with  $1/6$  chance.

Let  $Y : S \rightarrow \mathbb{R}$ . Notice that, if the child successfully matches 2 pictures with their words, then the third picture will be also successfully matched. Therefore, the child may correctly identify 0, 1, or 3 of the pictures presented and  $\text{Im } Y = \{0, 1, 3\}$ . Then, we have  $p_Y(y) = 0$  for any  $y \neq 0, 1, 3$ . For the other values of  $y$ :

- $p_Y(0) = P(Y = 0) = P(\{312, 231\}) = 1/3$ .
- $p_Y(1) = P(Y = 1) = P(\{132, 213, 312\}) = \frac{1}{2}$ .
- $p_Y(3) = P(Y = 3) = P(\{123\}) = \frac{1}{6}$ .

**PROBLEM 7.** The sample space is all distinct subsets of two numbers from  $\{1, 2, 3, 4, 5\}$ . There are  $\binom{5}{2} = 10$  possible outcomes and all of the outcome are equally likely to occur.

- a) We have  $\text{Im } X = \{2, 3, 4, 5\}$ . The number 1 is missing because in the two balls selected, if ball #1 is selected, then the other ball's number is automatically one of 2, 3, 4, 5. We will present the pmf of  $X$  in a table.

$x$	2	3	4	5
$p_X(x)$	1/10	1/5	3/10	2/5

To compute  $p_X(2)$ , we first notice that  $\{X = 2\} = \{\{1, 2\}\}$  and therefore  $P(X = 2) = 1/10$ . To compute  $p_X(3)$ , we first notice that  $\{X = 3\} = \{\{1, 3\}, \{2, 3\}\}$  and therefore  $P(X = 3) = 2/10 = 1/5$ . Similar calculations lead to the values of  $p_X(4)$  and  $p_X(5)$ . Notice that  $1/10 + 1/5 + 3/10 + 2/5 = 1$ .

- b) Removing the parenthesis in the set and considering them as unordered list, the outcomes of  $S$  can be explicitly enumerated:

$$S = \{12, 13, 14, 15, 23, 24, 25, 34, 35, 45\}.$$

Therefore, considering all the outcomes and adding the numbers, we see that  $\text{Im } X = \{3, 4, 5, 6, 7, 8, 9\}$ . We have, more precisely,  $X(12) = 3$ ,  $X(13) = 4$ ,  $X(14) = X(23) = 5$ ,  $X(24) = X(15) = 6$ ,  $X(25) = X(34) = 7$ ,  $X(35) = 8$ ,  $X(45) = 9$ . Using the same strategy as in a), we find the following values for  $p_X$ .

$x$	3	4	5	6	7	8	9
$p_X(x)$	1/10	1/10	1/5	1/5	1/5	1/10	1/10

**PROBLEM 8.** If  $p$  is a probability mass function, then we know it should satisfy  $\sum_{k=1}^{\infty} p(k) = 1$ . This gives the following condition:

$$\sum_{k=1}^{\infty} \frac{c}{k(k+1)} = 1.$$

Now, using the trick  $\frac{1}{k(k+1)} = \frac{1}{k} - \frac{1}{k+1}$ , we see that the series  $\sum_{k=1}^{\infty} \frac{1}{k(k+1)}$  is convergent and

$$\sum_{k=1}^{\infty} \frac{1}{k(k+1)} = \lim_{N \rightarrow \infty} \sum_{k=1}^N \left( \frac{1}{k} - \frac{1}{k+1} \right) = \lim_{N \rightarrow \infty} 1 - \frac{1}{N+1} = 1.$$

Therefore, using the properties of series, we see that

$$\sum_{k=1}^{\infty} \frac{c}{k(k+1)} = 1 \iff c \sum_{k=1}^{\infty} \frac{1}{k(k+1)} = 1 \iff c \cdot 1 = 1 \iff c = 1.$$

This means the function  $p$  is a pmf if and only if  $c = 1$ .

### C.3 Functions of Discrete Random Variables

**PROBLEM 9.** Setting  $X = 1$ ,  $X = 2$ ,  $X = 3$ , and  $X = 4$  in the expression of  $Y$ , we get  $Y = 0, 3, 8, 15$ . Therefore,  $\text{Im } Y = \{0, 3, 8, 15\}$ .

Using Theorem 3, with  $g(x) = x^2 - 1$ , we have

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

For

- $y = 0$ , we have  $g^{-1}(0) = \{x \in \text{Im } X : g(x) = 0\} = \{1\}$ ;
- $y = 3$ , we have  $g^{-1}(3) = \{x \in \text{Im } X : g(x) = 3\} = \{2\}$ ;
- $y = 8$ , we have  $g^{-1}(8) = \{x \in \text{Im } X : g(x) = 8\} = \{3\}$ ;
- $y = 15$ , we have  $g^{-1}(15) = \{x \in \text{Im } X : g(x) = 15\} = \{4\}$ .

Therefore,

- $p_Y(0) = P(X = 1) = 0.4$ .
- $p_Y(3) = P(X = 2) = 0.3$ .
- $p_Y(8) = P(X = 3) = 0.2$ .
- $p_Y(15) = P(X = 4) = 0.1$ .
- $p_Y(y) = 0$  for any other values  $y$  different from 0, 3, 8, 15.  $\triangle$

**PROBLEM 10.** Setting  $X = 1, 2, 3, 4$  in the expression of  $Y$ , we get  $Y = 1, 0, -1, 0$  respectively. Therefore,  $\text{Im } Y = \{-1, 0, 1\}$ .

Using Theorem 3 again, but with  $g(x) = \sin(\frac{\pi}{2}x)$ , we have

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

For

- $y = -1, g^{-1}(-1) = \{3\};$
- $y = 0, g^{-1}(0) = \{2, 4\};$
- $y = 1, g^{-1}(1) = \{1\}.$

Therefore,

- $p_Y(-1) = P(X = 3) = 0.2$ .
- $p_Y(0) = P(X = 2) + P(X = 4) = 0.3 + 0.1 = 0.4$ .
- $p_Y(1) = P(X = 1) = 0.4$ .  $\triangle$

## C.4 Expectation and Variance

**PROBLEM 11.** Let  $t_1$  be the label “the dimensions of the trailer are  $8 \times 10 \times 30$ ”. and let  $t_2$  be the label “the dimensions of the trailer are  $8 \times 10 \times 40$ ”. Given a trailer, the possible outcome is a trailer of type  $t_1$  or of type  $t_2$ . Therefore,  $S = \{t_1, t_2\}$  with  $P(\{t_1\}) = 0.3$  and  $P(\{t_2\}) = 0.7$ .

Let  $X$  be the map giving the volume of a trailer. We have

$$X(t_1) = 8 \cdot 10 \cdot 30 = 2400 \quad \text{and} \quad X(t_2) = 8 \cdot 10 \cdot 40 = 3200.$$

Therefore, we get

$$\text{Exp}(X) = X(t_1)P(X = 2400) + X(t_2)P(X = 3200) = (2400)(0.3) + (3200)(0.7) = 2960.$$

The average volume shipped per trailer load is 2960ft<sup>3</sup>.  $\triangle$

**PROBLEM 12.** A firm can be assigned one or two contracts. Therefore, we can generate the set of outcomes as couple of letters taken from  $\{a, b, c\}$ . For example  $aa$  means  $a$  was assigned to the two contracts, but  $ab$  or  $ba$  means that  $a$  and  $b$  was assigned to one of the contracts. The sample space  $S$  is

$$S = \{aa, ab, ba, ac, ca, bb, bc, cb, cc\}.$$

Since the firms are assigned a contract at random, each outcome are equally likely to occur, so with  $1/9$ .

a) In the first scenario, assume that  $X$  is the possible profit made by firm  $A$  after the contracts were assigned. Therefore, this means

- $X(aa) = 180,000$ .
- $X(ab) = X(ba) = X(ac) = X(ca) = 90,000$ .
- $X(bb) = X(bc) = X(cb) = X(cc) = 0$ .

The expectation is then calculated as followed:

$$\begin{aligned}
 \text{Exp}(X) &= 180,000P(X = 180,000) + 90,000P(X = 90,000) + 0P(X = 0) \\
 &= 180,000P(\{aa\}) + 90,000(P(\{ab, ba, ac, ca\}) + 0) \\
 &= \frac{180,000}{9} + \frac{90,000 \cdot 4}{9} \\
 &= 20,000 + 40,000 \\
 &= 60,000
 \end{aligned}$$

b) Let  $Y$  be the profit made by firms  $A$  and  $B$  after the contrasts were assigned. Therefore, this means

- $X(aa) = X(bb) = X(ab) = X(ba) = 180,000$ .
- $X(ac) = X(ca) = X(bc) = X(cb) = 90,000$ .
- $X(cc) = 0$ .

The expectation is then calculated as followed:

$$\begin{aligned}
 \text{Exp}(X) &= 180,000P(X = 180,000) + 90,000P(X = 90,000) + 0P(X = 0) \\
 &= 180,000P(\{aa, bb, ab, ba\}) + 90,000P(\{ac, ca, bc, cb\}) \\
 &= \frac{(180,000)(4)}{9} + \frac{(90,000)(4)}{9} \\
 &= 80,000 + 40,000 \\
 &= 120,000.
 \end{aligned}$$

△

**PROBLEM 13.** We have  $\text{Im } X = \{1, 2, 3, 4, 5, 6\}$  and each value of  $X$  has a chance of  $1/6$  to occur. Therefore,

$$\text{Exp}(X) = \frac{1}{6} + \frac{2}{6} + \frac{3}{6} + \frac{4}{6} + \frac{5}{6} + \frac{6}{6} = \frac{21}{6} = 3\frac{1}{2}.$$

The variance is calculated using formula in the Theorem C.6. We first have  $\text{Im } X^2 = \{1, 4, 9, 16, 25, 36\}$  and

$$\text{Exp}(X^2) = \frac{1}{6} + \frac{4}{6} + \frac{9}{6} + \frac{16}{6} + \frac{25}{6} + \frac{36}{6} = \frac{91}{6} = 15\frac{1}{6}.$$

Therefore,

$$\text{Var}(X) = \text{Exp}(X^2) - (\text{Exp}(X))^2 = \frac{91}{6} - \left(\frac{21}{6}\right)^2 = \frac{70}{6} = 11\frac{2}{3}.$$

Thus,

$$\sigma = \sqrt{\text{Var}(X)} = \sqrt{70/6} \approx 3.4157.$$

△

**PROBLEM 14.** By the formula in Theorem C.6, we have

$$\text{Var}(aX + b) = \text{Exp}((aX + b)^2) - (\text{Exp}(aX + b))^2.$$

We have  $(aX + b)^2 = a^2X^2 + 2abX + b^2$  and  $\text{Exp}(aX + b) = a\text{Exp}(X) + b$ . Therefore,

$$\begin{aligned}\text{Var}(aX + b) &= a^2\text{Exp}(X^2) + 2ab\text{Exp}(X) + b^2 - a^2(\text{Exp}(X))^2 - 2ab\text{Exp}(X) - b^2 \\ &= a^2\text{Exp}(X^2) - a^2\text{Exp}(X)^2 \\ &= a^2\text{Var}(X).\end{aligned}$$

△

## C.5 Conditional Expectation and the Partition Theorem

**PROBLEM 15.** Let  $B = \{X = x\}$ . Then, we have

$$E(g(X)|B) = \sum_{y \in \text{Im } g(X)} yP(g(X) = y|B).$$

However, if  $B$  has occurred, then  $X = x$  and the sum over  $\text{Im } g(X)$  is restricted to the value  $y = g(x)$ . Hence,

$$E(g(X)|B) = g(x)P(g(X) = g(x)|B) = g(x) \frac{P(\{g(X) = g(x)\} \cap \{X = x\})}{P(X = x)}.$$

But  $\{g(X) = g(x)\} = \{X = x\}$  and therefore

$$E(g(X)|B) = g(x) \frac{P(X = x)}{P(X = x)} = g(x).$$

△

## C.6 Examples of Discrete Random Variables

**PROBLEM 16.** The map  $X$  has a discrete range, that is  $\text{Im } X = \{0, 1, 2, 3, \dots, 30\}$ . However it does not have a binomial distribution because the probability that there is rain on a given day varies from day to day. Therefore, the parameter  $p$  is not fixed. △

**PROBLEM 17.**

- In this case, if it was explicitly mentioned “The number of students in a sample of  $X$  students who took the SAT”, then we could model the distribution of  $X$  on the binomial distribution with  $n = 100$  and  $q = 0.45$ . Unfortunately, it is not mentioned and therefore we can’t model the distribution with a binomial distribution.
- It can’t be model by a binomial distribution because there is not enough information to find the parameter  $q$ . We will see later that the distribution of the scores of the 100 students can be model by a normal distribution.
- If  $X_j$  is the random variable “The student labeled  $j$  scored above average on the SAT”, then  $X$ : “the number of students in the sample who scored above average on the SAT”, which is equal to  $X_1 + X_2 + \dots + X_{100}$ , has a binomial distribution. In this case,  $n = 100$  and the value of  $q$  is not possible to find. We would need more information to compute an approximate value for  $q$ . For example, with the additional assumption that the distribution of the student’s scores is a Normal distribution, then we can assume that  $q = 0.5$ , because  $P(X > \mu) = 0.5$  for any normal distribution.

- d) It can't be model by a binomial distribution because there is not enough information to find the parameter  $q$ . We would need additional information on the average time of a student to complete the test. We will see later that the distribution of the random variable will be modeled by a Normal distribution.  $\triangle$

**PROBLEM 18.** By Definition C.3, we have

$$\text{Exp}(X) = \sum_{k=0}^n kP(X = k) = \sum_{k=0}^n k \frac{n!}{k!(n-k)!} q^k (1-q)^{n-k}.$$

The term with  $k = 0$  disappears and we enter into the following chain of equalities:

$$\begin{aligned} \text{Exp}(X) &= \sum_{k=1}^n \frac{kn!}{k!(n-k)!} q^k (1-q)^{n-k} \\ &= \sum_{k=0}^{n-1} \frac{(k+1)n!}{(k+1)!(n-k-1)!} q^{k+1} (1-q)^{n-k-1} \\ &= nq \sum_{k=0}^{n-1} \frac{(n-1)!}{k!(n-1-k)!} q^k (1-q)^{n-k-1} \\ &= nq(q + (1-q))^{n-1} \\ &= nq. \end{aligned}$$

To compute the  $\text{Var}(X)$ , we use the formula in Theorem C.6. We have  $\text{Exp}(X) = nq$  from the previous calculations. We need to compute  $\text{Exp}(X^2)$ . By the Theorem C.4 we have

$$\text{Exp}(X^2) = \sum_{k=0}^n k^2 P(X^2 = k^2) = \sum_{k=0}^n k^2 P(X = k),$$

where  $\{X^2 = k^2\} = \{X = k\}$  because  $X$  assumes only non-negative integer values. Therefore,

$$\begin{aligned} \text{Exp}(X^2) &= \sum_{k=0}^n \frac{k^2 n!}{k!(n-k)!} q^k (1-q)^{n-k} \\ &= \sum_{k=1}^n \frac{k^2 n!}{k!(n-k)!} q^k (1-q)^{n-k} \\ &= nq \sum_{k=0}^{n-1} \frac{(k+1)(n-1)!}{k!(n-1-k)!} q^k (1-q)^{n-1-k} \\ &= nq \left( \sum_{k=0}^{n-1} \frac{k(n-1)!}{k!(n-1-k)!} q^k (1-q)^{n-1-k} + \sum_{k=0}^{n-1} \frac{(n-1)!}{k!(n-1-k)!} q^k (1-q)^{n-1-k} \right) \\ &= nq \left( \sum_{k=1}^{n-1} \frac{k(n-1)!}{k!(n-1-k)!} q^k (1-q)^{n-1-k} + 1 \right) \\ &= nq \left( (n-1)q \sum_{k=0}^{n-2} \frac{(n-2)!}{k!(n-2-k)!} q^k (1-q)^{n-2-k} + 1 \right) \\ &= nq \left( (n-1)q + 1 \right) \\ &= nq(nq - q + 1) \\ &= n^2 q^2 + nq(1-q). \end{aligned}$$



Hence,

$$\text{Var}(X) = n^2 q^2 + nq(1 - q) - n^2 q^2 = nq(1 - q). \quad \triangle$$