

# Problem Section 6

## Expected Value and Variance

### Learning Outcomes

The problems are designed to build conceptual understanding and problem-solving skills. The emphasis is on learning to find, evaluate and build confidence. The specific tasks include:

- Calculate expected value and variance using the PMF
- Use Chebychev's inequality to make probability statements
- Back up and support work with relevant explanations

### Exercises

1. (Law of the Unconscious Probabilist) Weird Wally offers you the following choice. You may have a flat amount of  $\frac{1}{3.5}$  dollars or you may roll a fair die and he will give you  $\frac{1}{X}$  dollars where  $X$  is the value on the roll. The PMF of  $X$  is shown in the table below.

$x$	1	2	3	4	5	6
$f(x)$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$

Which is the better deal? Calculate  $E\left[\frac{1}{X}\right]$  to decide. (Hint: use the Law of the Unconscious Probabilist - Lemma 7.2)

We can find  $E\left[\frac{1}{X}\right]$  as follows by using the Law of the Unconscious Probabilist:

$$\begin{aligned} E\left[\frac{1}{X}\right] &= \sum_x \frac{1}{x} \cdot f(x) \\ &= \frac{1}{6} \sum_{x=1}^6 \frac{1}{x} \end{aligned}$$

We may calculate this using R.

```
inv_x <- 1/(1:6)
E_1_x <- sum(inv_x)/6
E_1_x
```

```
## [1] 0.4083333
```

Using R we see that  $E\left[\frac{1}{X}\right] = 0.4083333$  versus Wally's offer of 0.2857143. Thus rolling the die has a better expected value, and this is the better deal.

2. (Mean and variance)

$x$	-\$5	\$170
$f(x)$	37/38	1/38

- a. Suppose a random variable  $X$  has PMF as shown below. Find  $E[X]$ . Also calculate  $SD[X]$ .

We may solve for  $E(X)$  and  $Var(X)$  using the definitions of these quantities:

$$E(X) = \sum_x xf(x) = -5 \times \frac{37}{38} + 170 \times \frac{1}{38}.$$

Since  $Var(X) = E(X^2) - E(X)^2$ . We know  $E(X)$ , so we can now find  $E(X^2)$

$$E(X^2) = \sum_x x^2 f(x) = -5^2 \times \frac{37}{38} + 170^2 \times \frac{1}{38}$$

These are calculated below.

```
E_X <- -5*37/38 + 170/38
Var_X <- (25*37/38 + 170^2/38 - E_X^2)
SD_X <- sqrt(Var_X)      #SD = sqrt(Var(X))
```

Thus we have that  $E(X) = -\$0.39$  and  $SD(X) = \$28.01$ .

The SD tells us the typical deviation in earnings from the average earnings ( $E(X)$ ). It is really high in this example. It tells us that typically our net gain will be in the interval  $[E(X) - SD(X), E(X) + SD(X)] = [-28.41, 27.62]$

- b. If  $X$  denotes a temperature of a randomly selected day recorded in degrees Fahrenheit, then  $Y = \frac{5}{9}X - \frac{160}{9}$  is the corresponding temperature in degrees Celsius. If the standard deviation for  $X$  is  $15.7^\circ F$ , what is the standard deviation of  $Y$ ?

By Lemma 7.4, we have the result that

$$Var(Y) = \left(\frac{5}{9}\right)^2 Var(X).$$

Since  $Var(X) = 15.7^2$  we can say that

$$Var(Y) = \frac{25 \times 15.7 \times 15.7}{81}$$

and

$$SD(Y) = \sqrt{Var(Y)} = \frac{5 \times 15.7}{9}.$$

3. (Chebychev) Suppose  $X$  is a random variable with mean and variance both equal to 20. What can be said about  $P(0 < X < 40)$ ?

Hint: Chebychev's inequality says that

$$P(|X - 20| \geq k\sqrt{20}) \leq \frac{1}{k^2}.$$

What would you choose for  $k$  here so you can say something about  $P(0 < X < 40)$ ?

If we choose  $k = \sqrt{20}$  then by Chebychev's inequality, we can say

$$P(|X - 20| \geq 20) \leq \frac{1}{20}. \tag{1}$$

Now it is important to note that the event  $0 < X < 40$  is actually the complement of the event  $|X - 20| \geq 20$  since

$$\begin{aligned} 0 < X < 40 &= (0 - 20) < X - 20 < (40 - 20), \\ &= -20 < X - 20 < 20, \\ &= |X - 20| < 20 \end{aligned}$$

Therefore

$$\begin{aligned} P(0 < X < 40) &= 1 - P(|X - 20| \geq 20), \\ &\geq 1 - \frac{1}{20} \end{aligned}$$

where the last inequality follows by reversing the one in (1)

4. Suppose we wish to generate  $X \sim \text{Binom}(n = 10, \pi = \frac{2}{3})$  subject to the constraint  $X \leq 3$ . Say we use the following naive algorithm to accomplish this task:

- Generate an  $x$  from a  $\text{Binom}(10, \frac{2}{3})$
- Accept the value  $x$  if  $x \leq 3$ . Otherwise reject it.

- a. Calculate the acceptance probability. That is, what is the probability we will accept a value  $x$  that is generated?

We accept  $x$  if  $x \leq 3$ . Thus the acceptance probability is  $P(X \leq 3)$ . We can find this using `pbinom`

```
pi <- pbinom(q = 3, size = 10, prob = 2/3)
```

The acceptance probability is 0.0197.

- b. Put your calculation from part a. to the test by generating 1,000 values from a  $\text{Binom}(10, \frac{2}{3})$  and calculate how often you would generate an acceptable value. Bump up the number of simulations to 5,000. What do you notice?

```
set.seed(131)

nrep <- 1000
x <- rbinom(n = nrep, size = 10, prob = 2/3)

sum(x <= 3)/nrep      #fraction of accepted values
```

```
## [1] 0.015
```

- c. Define a new random variable  $Y$  as the number of times we have to generate a binomial variable before we find an acceptable one. For example, if on our first try, we get an acceptable  $x$ , then  $y = 0$ . Write the PMF of  $Y$ .

The random variable  $Y$  represents the number of failures before we find an acceptable value. It therefore takes values  $0, 1, 2, \dots$

If  $Y = 0$ , then it means there were 0 failures and we accepted the first value generated. Therefore  $P(Y = 0) = \pi$  where  $\pi$  is the acceptance probability we found in part a.

If  $Y = 1$ , then it means there was 1 failure and then we generated an acceptable value. Therefore  $P(Y = 1) = (1 - \pi) \times \pi$ . We are of course relying on the fact that the outcomes of each random generation from the binomial PMF are independent of each other.

Similarly if  $Y = 2$ , then it means there were 2 failures and then we generated an acceptable value. Therefore  $P(Y = 2) = (1 - \pi)^2 \times \pi$ .

Hence the pattern is as follows.

$$f(y) = (1 - \pi)^y \times \pi, \quad y = 0, 1, 2, \dots$$