#### Section 2.4

# Cumulative Distribution Function (CDF)

#### **Cumulative Distribution**

### Definition 2.11 Function (CDF)

The cumulative distribution function (CDF) of random variable X is

$$F_X(x) = P[X \le x].$$

For any discrete random variable X with range  $S_X = \{x_1, x_2, \ldots\}$  satisfying  $x_1 \leq x_2 \leq \ldots$ ,

- (a)  $F_X(-\infty) = 0$  and  $F_X(\infty) = 1$ .
- (b) For all  $x' \ge x$ ,  $F_X(x') \ge F_X(x)$ .
- (c) For  $x_i \in S_X$  and  $\epsilon$ , an arbitrarily small positive number,

$$F_X(x_i) - F_X(x_i - \epsilon) = P_X(x_i).$$

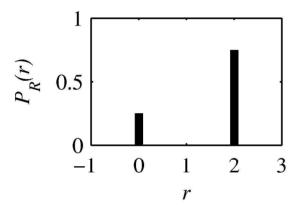
(d)  $F_X(x) = F_X(x_i)$  for all x such that  $x_i \le x < x_{i+1}$ .

For all  $b \ge a$ ,

$$F_X(b) - F_X(a) = P[a < X \le b].$$

#### **Example 2.23** Problem

In Example 2.6, we found that random variable R has PMF

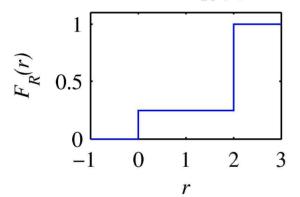


$$P_R(r) = \left\{ egin{array}{ll} 1/4 & r = 0, \\ 3/4 & r = 2, \\ 0 & ext{otherwise.} \end{array} 
ight.$$

Find and sketch the CDF of random variable R.

#### **Example 2.23** Solution

From the PMF  $P_R(r)$ , random variable R has CDF



$$F_R(r) = P[R \le r] = \begin{cases} 0 & r < 0, \\ 1/4 & 0 \le r < 2, \\ 1 & r \ge 2. \end{cases}$$

Keep in mind that at the discontinuities r=0 and r=2, the values of  $F_R(r)$  are the upper values:  $F_R(0)=1/4$ , and  $F_R(2)=1$ . Math texts call this the right hand limit of  $F_R(r)$ .

#### Section 2.5

# Averages

#### Definition 2.12 Mode

A mode of random variable X is a number  $x_{\text{mod}}$  satisfying  $P_X(x_{\text{mod}}) \ge P_X(x)$  for all x.

#### **Definition 2.13 Median**

A median,  $x_{\text{med}}$ , of random variable X is a number that satisfies

$$P\left[X < x_{\text{med}}\right] = P\left[X > x_{\text{med}}\right]$$

#### Definition 2.14 Expected Value

The expected value of X is

$$E[X] = \mu_X = \sum_{x \in S_X} x P_X(x).$$

This is also called as "mean."

#### **Example 2.25** Problem

For one quiz, 10 students have the following grades (on a scale of 0 to 10):

Find the mean, the median, and the mode.

#### **Example 2.25** Solution

The sum of the ten grades is 68. The mean value is 68/10 = 6.8. The median is 7 since there are four scores below 7 and four scores above 7. The mode is 5 since that score occurs more often than any other. It occurs three times.

The Bernoulli (p) random variable X has expected value E[X] = p.

#### **Proof: Theorem 2.4**

$$E[X] = 0 \cdot P_X(0) + 1P_X(1) = 0(1-p) + 1(p) = p.$$

#### **Example 2.26** Problem

Random variable R in Example 2.6 has PMF

$$P_R(r) = \begin{cases} 1/4 & r = 0, \\ 3/4 & r = 2, \\ 0 & \text{otherwise.} \end{cases}$$

What is E[R]?

# **Example 2.26 Solution**

$$E[R] = \mu_R = 0 \cdot P_R(0) + 2P_R(2) = 0(1/4) + 2(3/4) = 3/2.$$

The geometric (p) random variable X has expected value E[X] = 1/p.

#### **Proof: Theorem 2.5**

Let q = 1 - p. The PMF of X becomes

$$P_X(x) = \begin{cases} pq^{x-1} & x = 1, 2, \dots \\ 0 & \text{otherwise.} \end{cases}$$

The expected value E[X] is the infinite sum

$$E[X] = \sum_{x=1}^{\infty} x P_X(x) = \sum_{x=1}^{\infty} x p q^{x-1}.$$

Applying the identity of Math Fact B.7, we have

$$E[X] = p \sum_{x=1}^{\infty} xq^{x-1} = \frac{p}{q} \sum_{x=1}^{\infty} xq^x = \frac{p}{q} \frac{q}{1 - q^2} = \frac{p}{p^2} = \frac{1}{p}.$$

Math Fact B.7: If 
$$|q| < 1$$
,  $\sum_{i=1}^{\infty} iq^i = \frac{q}{(1-q)^2}$ .  $\Rightarrow E[X] = \frac{p}{q} \frac{q}{1-q^2} \Rightarrow \frac{p}{q} \frac{q}{(1-q)^2}$ 

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The Poisson  $(\alpha)$  random variable in Definition 2.10 has expected value  $E[X] = \alpha$ .

#### **Proof: Theorem 2.6**

$$E[X] = \sum_{x=0}^{\infty} x P_X(x) = \sum_{x=0}^{\infty} x \frac{\alpha^x}{x!} e^{-\alpha}.$$

We observe that x/x! = 1/(x-1)! and also that the x=0 term in the sum is zero. In addition, we substitute  $\alpha^x = \alpha \cdot \alpha^{x-1}$  to factor  $\alpha$  from the sum to obtain

$$E[X] = \alpha \sum_{x=1}^{\infty} \frac{\alpha^{x-1}}{(x-1)!} e^{-\alpha}.$$

Next we substitute l = x - 1, with the result

$$E[X] = \alpha \sum_{l=0}^{\infty} \frac{\alpha^{l}}{l!} e^{-\alpha} = \alpha.$$

We can conclude that the marked sum equals 1 either by invoking the identity  $e^{\alpha} = \sum_{l=0}^{\infty} \alpha^l / l!$  or by applying Theorem 2.1(b) to the fact that the marked sum is the sum of the Poisson PMF over all values in the range of the random variable.

#### Section 2.6

# Functions of a Random Variable

#### Definition 2.15 Derived Random Variable

Each sample value y of a derived random variable Y is a mathematical function g(x) of a sample value x of another random variable X. We adopt the notation Y = g(X) to describe the relationship of the two random variables.

#### **Example 2.27** Problem

The random variable X is the number of pages in a facsimile transmission. Based on experience, you have a probability model  $P_X(x)$  for the number of pages in each fax you send. The phone company offers you a new charging plan for faxes: \$0.10 for the first page, \$0.09 for the second page, etc., down to \$0.06 for the fifth page. For all faxes between 6 and 10 pages, the phone company will charge \$0.50 per fax. (It will not accept faxes longer than ten pages.) Find a function Y = g(X) for the charge in cents for sending one fax.

#### **Example 2.27 Solution**

The following function corresponds to the new charging plan.

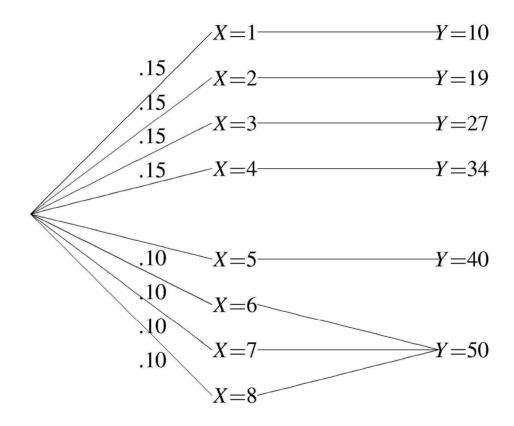
$$Y = g(X) = \begin{cases} 10.5X - 0.5X^2 & 1 \le X \le 5\\ 50 & 6 \le X \le 10 \end{cases}$$

You would like a probability model  $P_Y(y)$  for your phone bill under the new charging plan. You can analyze this model to decide whether to accept the new plan.

For a discrete random variable X, the PMF of Y = g(X) is

$$P_{Y}(y) = \sum_{x:g(x)=y} P_{X}(x).$$

#### Figure 2.1



The derived random variable Y = g(X) for Example 2.29. Yates Chapter 2

#### **Example 2.28** Problem

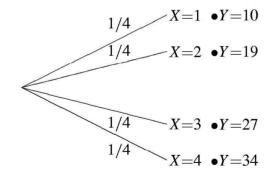
In Example 2.27, suppose all your faxes contain 1, 2, 3, or 4 pages with equal probability. Find the PMF and expected value of Y, the charge for a fax.

#### **Example 2.28** Solution

From the problem statement, the number of pages *X* has PMF

$$P_X(x) = \begin{cases} 1/4 & x = 1, 2, 3, 4, \\ 0 & \text{otherwise.} \end{cases}$$

The charge for the fax, Y, has range  $S_Y = \{10, 19, 27, 34\}$  corresponding to  $S_X = \{1, 2, 3, 4\}$ . The experiment can be described by the following tree. Here each value of Y results in a unique value of X. Hence, we can use Equation (2.66) to find  $P_Y(y)$ .

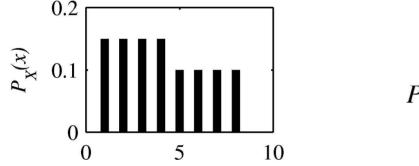


$$P_Y(y) = \begin{cases} 1/4 & y = 10, 19, 27, 34, \\ 0 & \text{otherwise.} \end{cases}$$

The expected fax bill is E[Y] = (1/4)(10 + 19 + 27 + 34) = 22.5 cents.

#### **Example 2.29 Problem**

Suppose the probability model for the number of pages X of a fax in Example 2.28 is



$$P_X(x) = \begin{cases} 0.15 & x = 1, 2, 3, 4 \\ 0.1 & x = 5, 6, 7, 8 \\ 0 & \text{otherwise} \end{cases}$$

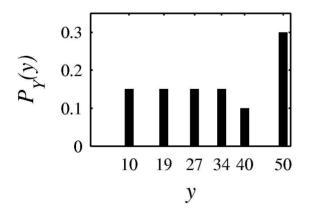
For the pricing plan given in Example 2.27, what is the PMF and expected value of Y, the cost of a fax?

#### **Example 2.29 Solution**

Now we have three values of X, specifically (6,7,8), transformed by  $g(\cdot)$  into Y=50. For this situation we need the more general view of the PMF of Y, given by Theorem 2.9. In particular,  $y_6=50$ , and we have to add the probabilities of the outcomes X=6, X=7, and X=8 to find  $P_Y(50)$ . That is,

$$P_Y(50) = P_X(6) + P_X(7) + P_X(8) = 0.30.$$

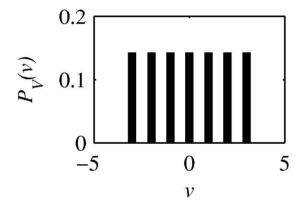
The steps in the procedure are illustrated in the diagram of Figure 2.1. Applying Theorem 2.9, we have



$$P_Y(y) = \begin{cases} 0.15 & y = 10, 19, 27, 34, \\ 0.10 & y = 40, \\ 0.30 & y = 50, \\ 0 & \text{otherwise.} \end{cases}$$

#### **Example 2.30** Problem

The amplitude V (volts) of a sinusoidal signal is a random variable with PMF

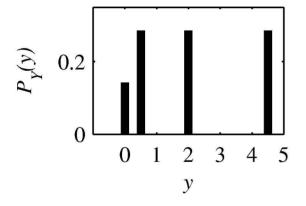


$$P_V(v) = \begin{cases} 1/7 & v = -3, -2, \dots, 3 \\ 0 & \text{otherwise} \end{cases}$$

Let  $Y = V^2/2$  watts denote the average power of the transmitted signal. Find  $P_Y(y)$ .

#### **Example 2.30 Solution**

The possible values of Y are  $S_Y = \{0, 0.5, 2, 4.5\}$ . Since Y = y when  $V = \sqrt{2y}$  or  $V = -\sqrt{2y}$ , we see that  $P_Y(0) = P_V(0) = 1/7$ . For y = 1/2, 2, 9/2,  $P_Y(y) = P_V(\sqrt{2y}) + P_V(-\sqrt{2y}) = 2/7$ . Therefore,



$$P_Y(y) = \begin{cases} 1/7 & y = 0, \\ 2/7 & y = 1/2, 2, 9/2, \\ 0 & \text{otherwise.} \end{cases}$$

#### **Quiz 2.6**

Monitor three phone calls and observe whether each one is a voice call or a data call. The random variable N is the number of voice calls. Assume N has PMF

$$P_N(n) = \begin{cases} 0.1 & n = 0, \\ 0.3 & n = 1, 2, 3, \\ 0 & \text{otherwise.} \end{cases}$$

Voice calls cost 25 cents each and data calls cost 40 cents each. T cents is the cost of the three telephone calls monitored in the experiment.

(1) Express T as a function of N. (2) Find  $P_T(t)$  and E[T].

#### Section 2.7

# Expected Value of a Derived Random Variable

Given a random variable X with PMF  $P_X(x)$  and the derived random variable Y = g(X), the expected value of Y is

$$E[Y] = \mu_Y = \sum_{x \in S_X} g(x) P_X(x)$$

#### **Proof: Theorem 2.10**

From the definition of E[Y] and Theorem 2.9, we can write

$$E[Y] = \sum_{y \in S_Y} y P_Y(y) = \sum_{y \in S_Y} y \sum_{x:g(x)=y} P_X(x) = \sum_{y \in S_Y} \sum_{x:g(x)=y} g(x) P_X(x),$$

where the last double summation follows because g(x) = y for each x in the inner sum. Since g(x) transforms each possible outcome  $x \in S_X$  to a value  $y \in S_Y$ , the preceding double summation can be written as a single sum over over all possible values  $x \in S_X$ . That is,

$$E[Y] = \sum_{x \in S_X} g(x) P_X(x)$$

#### **Example 2.31** Problem

In Example 2.28,

$$P_X(x) = \begin{cases} 1/4 & x = 1, 2, 3, 4, \\ 0 & \text{otherwise,} \end{cases}$$

and

$$Y = g(X) = \begin{cases} 10.5X - 0.5X^2 & 1 \le X \le 5, \\ 50 & 6 \le X \le 10. \end{cases}$$

What is E[Y]?

#### **Example 2.31 Solution**

Applying Theorem 2.10 we have

$$E[Y] = \sum_{x=1}^{4} P_X(x) g(x)$$

$$= (1/4)[(10.5)(1) - (0.5)(1)^2] + (1/4)[(10.5)(2) - (0.5)(2)^2]$$

$$+ (1/4)[(10.5)(3) - (0.5)(3)^2] + (1/4)[(10.5)(4) - (0.5)(4)^2]$$

$$= (1/4)[10 + 19 + 27 + 34] = 22.5 \text{ cents.}$$

For any random variable X,

$$E\left[X - \mu_X\right] = 0.$$

#### **Proof: Theorem 2.11**

Defining  $g(X) = X - \mu_X$  and applying Theorem 2.10 yields

$$E[g(X)] = \sum_{x \in S_X} (x - \mu_X) P_X(x) = \sum_{x \in S_X} x P_X(x) - \mu_X \sum_{x \in S_X} P_X(x).$$

The first term on the right side is  $\mu_X$  by definition. In the second term,  $\sum_{x \in S_X} P_X(x) = 1$ , so both terms on the right side are  $\mu_X$  and the difference is zero.

For any random variable X,

$$E[aX + b] = aE[X] + b.$$

#### **Example 2.32 Problem**

Recall that in Examples 2.6 and 2.26, we found that R has PMF

$$P_R(r) = \begin{cases} 1/4 & r = 0, \\ 3/4 & r = 2, \\ 0 & \text{otherwise,} \end{cases}$$

and expected value E[R] = 3/2. What is the expected value of V = g(R) = 4R + 7?

#### **Example 2.32 Solution**

From Theorem 2.12,

$$E[V] = E[g(R)] = 4E[R] + 7 = 4(3/2) + 7 = 13.$$

We can verify this result by applying Theorem 2.10. Using the PMF  $P_R(r)$  given in Example 2.6, we can write

$$E[V] = g(0)P_R(0) + g(2)P_R(2) = 7(1/4) + 15(3/4) = 13.$$

# **Example 2.33 Problem**

In Example 2.32, let  $W = h(R) = R^2$ . What is E[W]?

### **Example 2.33 Solution**

Theorem 2.10 gives

$$E[W] = \sum h(r)P_R(r) = (1/4)0^2 + (3/4)2^2 = 3.$$

Note that this is not the same as  $h(E[W]) = (3/2)^2$ .