

# Chapter 2

## Random Variables and Probability Distributions

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## 2.1 Random Variables

### 2.1.1 Introduction to random Variables

- Let  $S$  be a sample space of an experiment.

As noted previously, the outcome of the experiment, or the points in  $S$ , need not be numbers.

For example, in tossing a coin the outcomes are  $H$  (heads) or  $T$  (tails), and in tossing a pair of dice the outcomes are pairs of integers.

- However, we frequently wish to assign a specific number to each outcome of the experiment.

For example, in coin tossing, it may be convenient to assign 1 to  $H$  and 0 to  $T$ ; or, in the tossing of a pair of dice, we may want to assign the sum of the two integers to the outcome.

- Such an assignment of numerical values is called a *random variable*.

More generally, we have the following definition.

## 2.1.2 Definition of a Random Variable

- Definition 2.1

A *random variable*  $X$  is a rule (function) that assigns (associates) a numerical value (a real number) to each outcome in a sample space  $S$  of the experiment.

We shall let  $R_X$  denote the set of numbers assigned by a random variable  $X$ , and we shall refer to  $R_X$  as the *range space*.

Clearly, a random variable is a function from the sample space  $S$  of the experiment to the real numbers  $\mathbb{R}$  and  $R_X$  is the range of  $X$ .

- We shall use a capital letter, say  $X$ , to denote a random variable and its corresponding small letter,  $x$  in this case, for one of its values.

## Example 2.1

Consider the experiment of tossing a pair of fair dice.

The sample space  $S$  consists of the 36 ordered pairs  $(a, b)$ , where  $a$  and  $b$  can be any of the integers from 1 to 6: that is,

$$S = \{(1,1), \dots, (1,6), (2,1), \dots, (2,6), (3,1), \dots, (3,6), (4,1), \dots, (4,6), \\ (5,1), \dots, (5,6), (6,1), \dots, (6,6)\}.$$

Let  $X$  assign to each point in  $S$  the sum of the numbers; then  $X$  is a random variable with range space  $R_X = \{2,3,4,5,6,7,8,9,10,11,12\}$ .

Let  $Y$  assign to each point the maximum of the two numbers; then  $Y$  is a random variable with range space  $R_Y = \{1,2,3,4,5,6\}$ .

Note that the random variable  $X$  assumes the value 8 for all elements in the subset  $E = \{(2,6), (3,5), (4,4), (5,3), (6,2)\}$  of the sample space  $S$ .

That is, each possible value of  $X$  represents an event that is a subset of the sample space for the given experiment.

## Example 2.2

The sample space of each possible outcome when three electronic components are tested may be written

$$S = \{NNN, NND, NDN, DNN, NDD, DND, DDN, DDD\},$$

where  $N$  denotes nondefective and  $D$  denotes defective.

Let  $X$  be the the number of defectives that occur.

Thus, each point in the sample space will be assigned a numerical value of 0, 1, 2, or 3.

Then  $X$  is a random variable with range space  $R_X = \{0,1,2,3\}$ .

Note that the random variable  $X$  assumes the value 2 for all elements in the subset  $E = \{DDN, DND, NDD\}$  of the sample space  $S$ .

That is, each possible value of  $X$  represents an event that is a subset of the sample space for the given experiment.

### Example 2.3

A box contains 12 items of which 3 are defective. A sample of three items is selected from the box.

The sample space  $S$  consists of  $\binom{12}{3} = 220$  different samples of size 3.

Let  $X$  denote the number of defective items in the sample; then  $X$  is a random variable with range space  $R_X = \{0,1,2,3\}$ .

## Example 2.4

Two balls are drawn in succession without replacement from an urn containing 4 red balls and 3 black balls.

For this experiment sample space  $S = \{RR, RB, BR, BB\}$ .

Let  $Y$  be the number of red balls drawn.

The possible outcomes and the values  $y$  of the random variable  $Y$ , are given in the following table:

Sample Space	$y$
$RR$	2
$RB$	1
$BR$	1
$BB$	0

### Example 2.5

Statisticians use sampling plans to either accept or reject batches or lots of material.

Suppose one of these sampling plans involves sampling independently 10 items from a lot of 100 items in which 12 are defective.

In this experiment the sample space consists of  $\binom{100}{10}$  different samples of size 10.

Let  $X$  be the random variable defined as the number of items found defective in the sample of 10.

In this case, the random variable  $X$  takes on the values 0, 1, 2, . . . , 9, 10.

### Example 2.6

Let  $X$  be the random variable defined by the waiting time, in hours, between successive speeders spotted by a radar unit.

The random variable  $X$  takes on all values  $x$  for which  $x \geq 0$ .



### Example 2.7

Consider the experiment of tossing a coin repeatedly until a head appears.

Let  $X$  be the random variable that the number of tosses required to get the first head, and

$Y$  be the random variable that the number of tosses required before getting the first head.

Then the sample space for the experiment  $S = \{H, TH, TTH, TTTH, \dots\}$  and has infinite number of elements.

Thus,  $R_X = \{1, 2, 3, \dots\}$  and  $R_Y = \{0, 1, 2, 3, \dots\}$ .

### 2.1.3 Discrete and Continuous Random Variables

- A random variable is called a *discrete random variable* if its set of possible outcomes is countable.
- In other words, random variables whose set of possible values can be written either as a finite sequence  $x_1, x_2, \dots, x_n$ , or as an infinite sequence  $x_1, x_2, \dots$  are said to be *discrete*.
- The random variables in Examples 2.1 to 2.4 are discrete random variables.
- There are random variables whose set of possible values is an entire interval of numbers on the real line, and hence are not discrete.
- When a random variable can take on values on a continuous scale (uncountable), it is called a *continuous random variable*.
- Often the possible values of a continuous random variable are precisely the same values that are contained in the continuous sample space.

- Obviously, the random variable described in Example 2.5 is a continuous random variable.
- Another example for a continuous random variable is the random variable denoting the lifetime of a car, when the car's lifetime is assumed to take on any value in some interval  $(a, b)$ .
- In most practical problems, continuous random variables represent *measured* data, such as all possible heights, weights, temperatures, distance, or life periods, whereas discrete random variables represent *count* data, such as the number of defectives in a sample of  $k$  items or the number of highway fatalities per year in a given city.

## Exercises

1. Classify the following random variables as discrete or continuous:

$X$ : the number of automobile accidents per year in Virginia.

$Y$ : the length of time to play 18 holes of golf.

$M$ : the amount of milk produced yearly by a particular cow.

$N$ : the number of eggs laid each month by a hen.

$P$ : the number of building permits issued each month in a certain city.

$Q$ : the weight of grain produced per acre.

2. An overseas shipment of 5 foreign automobiles contains 2 that have slight paint blemishes. If an agency receives 3 of these automobiles at random, list the elements of the sample space  $S$ , using the letters  $B$  and  $N$  for blemished and nonblemished, respectively; then to each sample point assign a value  $x$  of the random variable  $X$  representing the number of automobiles with paint blemishes purchased by the agency.

## 2.2 Discrete Probability Distributions

### 2.2.1 Probability Mass Function (pmf)

- A discrete random variable assumes each of its values with a certain probability.
- Let us consider the case of tossing a coin three times and let  $X$  be the random variable representing the number of heads.

For this experiment the sample space  $S$  consists of 8 elements and is given by

$$S = \{TTT, HTT, THT, TTH, HHT, HTH, THH, HHH\}.$$

The range space of the random variable  $X$  is  $R_X = \{0,1,2,3\}$ .

$X$  assumes the value 0 with probability  $1/8$ , since only 1 of the 8 equally likely sample points result in zero heads and three tails.

$X$  assumes the value 1 with probability  $3/8$ , since 3 of the 8 equally likely sample points result in one head and two tails.

$X$  assumes the value 2 with probability  $3/8$ , since 3 of the 8 equally likely sample points result in two heads and one tail.

$X$  assumes the value 3 with probability  $1/8$ , since only 1 of the 8 equally likely sample points result in 3 heads and zero tails.

Now we can summarize the values  $x$  of the random variable  $X$  together with the corresponding probabilities as in the following table:

$x$	0	1	2	3
$P(X = x)$	$1/8$	$3/8$	$3/8$	$1/8$

Notice that,  $P(X = 0) + P(X = 1) + P(X = 2) + P(X = 3) = 1$  and  $P(X = x) > 0$  for all  $x$  in  $R_X$ .

- Frequently, it is convenient to represent all the probabilities of a random variable  $X$  by a formula.
- Such a formula would necessarily be a function of the numerical values  $x$  that we shall denote by  $f(x)$  (or  $f_X(x)$ ).

- Definition 5 - Probability Mass Function of a Discrete Random Variable

Let  $X$  be a discrete random variable with distinct values  $x_1, x_2, \dots, x_n, \dots$ .

Then the function, denoted by  $f(x)$  (or  $f_X(x)$ ) and defined by

$$f(x) = \begin{cases} P(X = x_i), & \text{if } x = x_i, i = 1, 2, \dots, n, \dots \\ 0, & \text{if } x \neq x_i, \end{cases}$$

is called the *probability function* or *probability mass function* of the discrete random variable  $X$ .

The set of ordered pairs  $(x_1, f(x_1)), (x_2, f(x_2)), \dots, (x_n, f(x_n)), \dots$ , is called the *probability distribution* of the discrete random variable  $X$ , and it is usually given as a formula or as a table.

- Note that,  $f(x)$  (or  $f_X(x)$ ) is a function with domain the real line and codomain the interval  $[0,1]$ .
- A graph of  $f(x)$  (or  $f_X(x)$ ) is called a *probability graph*.

- In general, a *probability function* or *probability mass function* of a discrete random variable  $X$  satisfy the following two properties:

1.  $f(x) \geq 0$ ,
2.  $\sum_x f(x) = 1$ ,

for each possible outcome  $x$  of the discrete random variable  $X$ .

- In the case that, the sample space  $S$  is an equiprobable space, we can easily obtain the distribution of a random variable from the following result.

- **Theorem 2.1**

Let  $S$  be an equiprobable space, and let  $X$  be a random variable on  $S$  with range space  $R_X = \{x_1, x_2, \dots, x_n\}$ .

Then  $f(x_i) = P(X = x_i) = \frac{\text{number of points in } S \text{ whose image is } x_i}{\text{number of points in } S}$



## Example 2.8

Again consider the experiment of tossing of a pair of fair dice in Example 2.1.

Let  $X$  be the random variable which gives the sum of the numbers appear on the two dice.

There are 36 outcomes in the sample space and range space  $R_X = \{2, 3, \dots, 12\}$ .

There is only one outcome (1,1) whose sum is 2; hence  $P(X = 2) = \frac{1}{36}$ .

There are two outcomes, (1,2) and (2,1), whose sum is 3; hence  $P(X = 3) = \frac{2}{36}$ .

There are three outcomes, (1,3), (2,2) and (3,1), whose sum is 4; hence  $P(X = 4) = \frac{3}{36}$ .

Similarly,  $P(X = 5) = \frac{4}{36}, P(X = 6) = \frac{5}{36}, \dots, P(X = 12) = \frac{1}{36}$ .

The probability distribution of  $X$  consists of the points in  $R_X$  along with their respective probabilities given by the following table:

$x_i$	2	3	4	5	6	7	8	9	10	11	12
$P(X = x_i)$	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$

### Example 2.9

Again consider the experiment of selecting a sample of 3 items from a box containing 12 items of which 3 are defective in Example 2.3.

The sample space  $S$  consists of  $\binom{12}{3} = 220$  different samples of size 3.

Let  $X$  denote the number of defective items in the sample; then  $X$  is a random variable with range space  $R_X = \{0,1,2,3\}$ .

There are  $\binom{9}{3} = 84$  samples of size 3 with no defective items; hence  $P(x =$

There is only one sample of size 3 containing the three defective items; hence  $P(x = 3) = \frac{1}{220}$ .

Then the probability distribution of  $X$  is as follows:

$x_i$	0	1	2	3
$P(X = x_i)$	$\frac{84}{220}$	$\frac{108}{220}$	$\frac{27}{220}$	$\frac{1}{220}$

### Example 2.10

A shipment of 20 similar laptop computers to a retail outlet contains 3 that are defective. If a school makes a random purchase of 2 of these computers, find the probability distribution for the number of defectives.

Let  $X$  be a random variable whose values  $x$  are the possible numbers of defective computers purchased by the school.

Then  $x$  can only take the numbers 0, 1, and 2.

$$\text{Now } f(0) = P(X = 0) = \frac{\binom{3}{0}\binom{17}{2}}{\binom{20}{2}} = \frac{68}{95}, f(1) = P(X = 1) = \frac{\binom{3}{1}\binom{17}{1}}{\binom{20}{2}} = \frac{51}{190},$$

$$\text{and } f(2) = P(X = 2) = \frac{\binom{3}{2}\binom{17}{0}}{\binom{20}{2}} = \frac{3}{190}.$$

Thus the probability distribution of  $X$  is

$x_i$	0	1	2
$P(X = x_i)$	68/95	51/190	3/190

### 2.2.2 Cumulative Distribution Function (cdf)

- There are many problems where we may wish to compute the probability that the observed value of a random variable  $X$  will be less than or equal to some real number  $x$ .
- By writing  $F(x) = P(X \leq x)$  for every real number  $x$ , we define  $F(x)$  (or  $F_X(x)$ ) to be the *cumulative distribution function* of the random variable  $X$ .
- Definition 6 - Cumulative Distribution Function of a Discrete Random Variable

The *cumulative distribution function*  $F(x)$  (or  $F_X(x)$ ) of a discrete random variable  $X$  with probability distribution  $f(x)$  is

$$F(x) = P(X \leq x) = \sum_{t \leq x} f(t),$$

where  $x$  is any real number, i.e.,  $-\infty < x < \infty$ .

- The cumulative distribution function  $F(x)$  has the following properties:
  1.  $F(x)$  is nondecreasing [i.e.,  $F(x) \leq F(y)$  if  $x \leq y$ ].
  2.  $\lim_{x \rightarrow -\infty} F(x) = 0$  and  $\lim_{x \rightarrow \infty} F(x) = 1$ .
  3.  $F(x)$  is continuous from the right [i.e.,  $\lim_{h \rightarrow 0^+} F(x + h) = F(x)$  for all  $x$ ].
- Suppose that the discrete random variable  $X$  takes on only a finite number of values  $x_1, x_2, \dots, x_n$ . Then the cumulative distribution function  $F(x)$  is given by

$$F(x) = \begin{cases} 0 & -\infty < x < x_1 \\ f(x_1) & x_1 \leq x < x_2 \\ f(x_1) + f(x_2) & x_2 \leq x < x_3 \\ \vdots & \vdots \\ f(x_1) + f(x_2) + \dots + f(x_n) & x_n \leq x < \infty \end{cases}$$

### Example 2.11

Suppose that a fair coin is tossed twice so that the sample space is  $S = \{HH, HT, TH, TT\}$ .

Let  $X$  represent the number of heads that can come up.

Thus, range space  $R_X = \{0,1,2\}$ .

The probability density function corresponding to the random variable  $X$  is given by

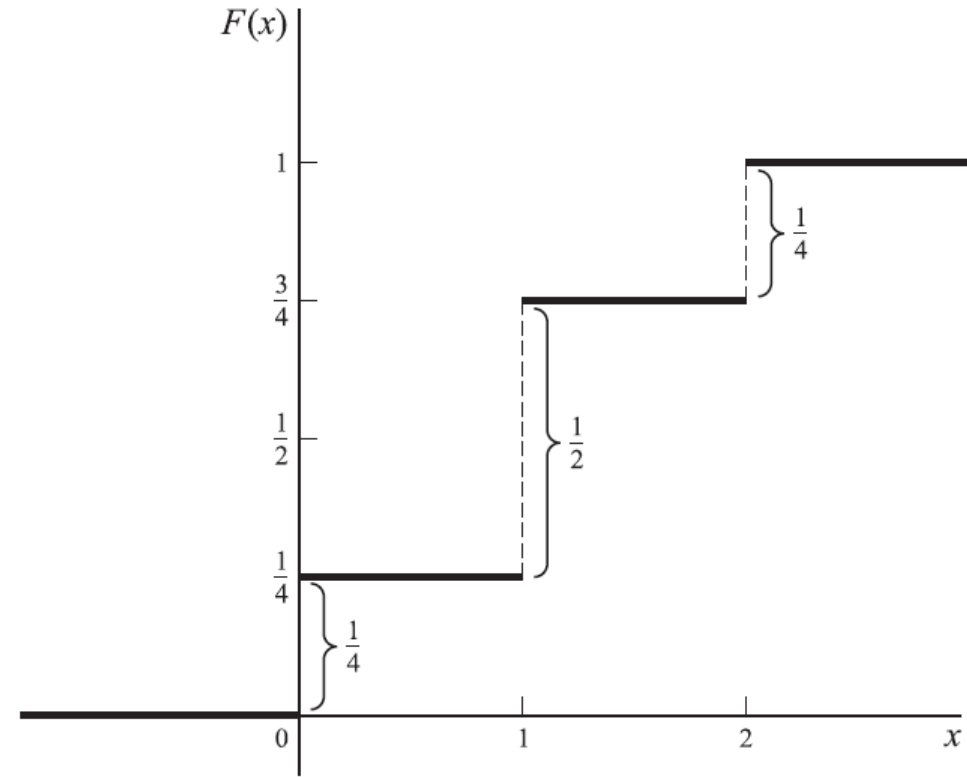
$x_i$	0	1	2
$P(X = x_i)$	$\frac{1}{4}$	$\frac{2}{4}$	$\frac{1}{4}$



The cumulative distribution function is given by

$$F(x) = \begin{cases} 0 & -\infty < x < 0 \\ 1/4 & 0 \leq x < 1 \\ 3/4 & 1 \leq x < 2 \\ 1 & 2 \leq x < \infty \end{cases}$$

The graph of  $F(x)$  is given below:



- One should pay particular notice to the fact that the cumulative distribution function is a monotone non-decreasing function defined not only for the values assumed by the given random variable but for all real numbers.

## Example 2.12

Again consider the experiment of tossing of a pair of fair dice discussed in Example 2.1 and Example 2.8.

Let  $X$  be the random variable which gives the sum of the numbers appear on the two dice.

As you know, there are 36 outcomes in the sample space and range space  $R_X = \{2, 3, \dots, 12\}$ .

From Example 2.8, the probability distribution of the random variable  $X$  is given by the following table:

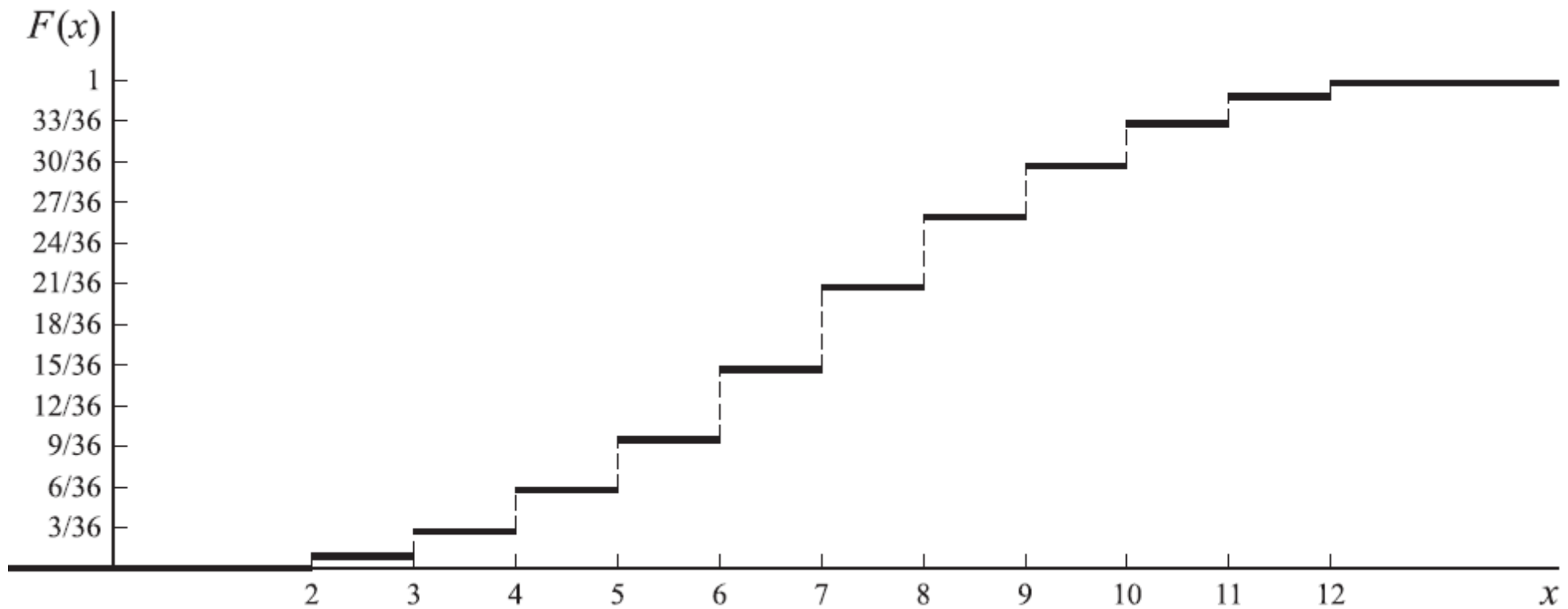
$x_i$	2	3	4	5	6	7	8	9	10	11	12
$P(X = x_i)$	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$

By using  $F(x) = P(X \leq x) = \sum_{t \leq x} f(t)$  and the probability distribution table we can compute the cumulative distribution function of the random variable  $X$ .

Thus,

$$F(x) = \begin{cases} 0, & -\infty < x < 2 \\ 1/36, & 2 \leq x < 3 \\ 3/36, & 3 \leq x < 4 \\ 6/36, & 4 \leq x < 5 \\ \vdots & \vdots \\ 35/36, & 11 \leq x < 12 \\ 1, & 12 \leq x < \infty \end{cases} .$$

The graph of the cumulative distribution function  $F(x)$  is given below.



### Example 2.13

If a car agency sells 50% of its inventory of a certain foreign car equipped with side airbags, find a formula for the probability distribution of the number of cars with side airbags among the next 4 cars sold by the agency.

Hence, obtain the cumulative distribution function for the above probability distribution.

Plot the probability distribution and the cumulative distribution function.

### Solution:

The  $2^4 = 16$  points in the sample space are equally likely to occur, since the probability of selling an automobile with side airbags is  $\frac{50}{100} = \frac{1}{2}$ .

The event of selling  $x$  models with side airbags and  $4 - x$  models without side airbags can occur in  $\binom{4}{x}$  ways, where  $x$  can be 0, 1, 2, 3, or 4.

Thus, the probability of selling  $x$  models with side airbags is given by

$$P(X = x) = \binom{4}{x} \left(\frac{1}{2}\right)^x \left(\frac{1}{2}\right)^{4-x} = \binom{4}{x} \left(\frac{1}{2}\right)^4 = \frac{1}{16} \binom{4}{x},$$

where  $x$  can be 0,1,2,3 or 4.

Let be  $X$  be the random variable of selling  $x$  models with side airbags.

Then the probability distribution of the random variable  $X$  is given by

$$f(x) = P(X = x) = \frac{1}{16} \binom{4}{x}, \text{ where } x \text{ can be } 0,1,2,3 \text{ or } 4.$$

From  $f(x) = \frac{1}{16} \binom{4}{x}$ , we get  $f(0) = \frac{1}{16}$ ,  $f(1) = \frac{1}{4}$ ,  $f(2) = \frac{3}{8}$ ,  $f(3) = \frac{1}{4}$ , and  $f(4) = \frac{1}{16}$ .

Therefore,  $F(0) = f(0) = \frac{1}{16}$ ,

$$F(1) = f(0) + f(1) = \frac{1}{16} + \frac{1}{4} = \frac{5}{16},$$

$$F(2) = f(0) + f(1) + f(2) = \frac{5}{16} + \frac{3}{8} = \frac{11}{16},$$

$$F(3) = f(0) + f(1) + f(2) + f(3) = \frac{11}{16} + \frac{1}{4} = \frac{15}{16},$$

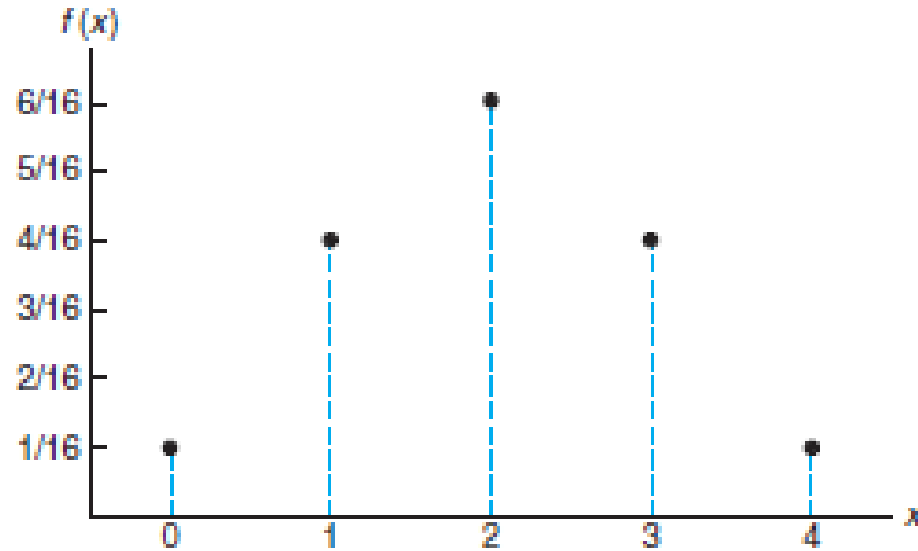
$$F(4) = f(0) + f(1) + f(2) + f(3) + f(4) = \frac{15}{16} + \frac{1}{16} = 1.$$

Hence,

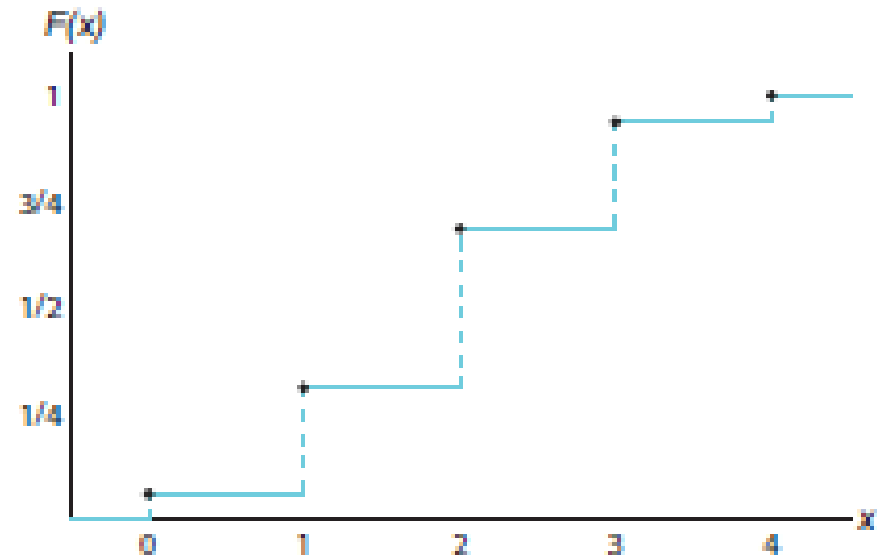
$$F(x) = \begin{cases} 0, & -\infty < x < 0, \\ \frac{1}{16}, & 0 \leq x < 1, \\ \frac{5}{16}, & 1 \leq x < 2, \\ \frac{11}{16}, & 2 \leq x < 3, \\ \frac{15}{16}, & 3 \leq x < 4, \\ 1, & 4 \leq x < \infty. \end{cases}.$$

Further we can observe that,  $f(2) = F(2) - F(1) = \frac{11}{16} - \frac{5}{16} = \frac{6}{16} = \frac{3}{8}.$

Plots of the probability distribution and the cumulative distribution function are given below:



Plot of probability distribution



Plot of cumulative distribution function



## Exercises

1. Determine the value  $c$  so that each of the following functions can serve as a probability distribution of the discrete random variable  $X$ :
  - (a)  $f(x) = c(x^2 + 4)$ , for  $x = 0, 1, 2, 3$ ;
  - (b)  $f(x) = c \binom{2}{x} \binom{3}{3-x}$ , for  $x = 0, 1, 2$ .
2. Find the probability distribution of boys and girls in families with 3 children, assuming equal probabilities for boys and girls.

If  $X$  represents the number of boys in the family, find the sample space  $S$ , range space  $R_X$  and the probability distribution of  $X$ .

Find the cumulative distribution function  $F(x)$  for the random variable  $X$  and plot this cumulative distribution function  $F(x)$ .

3. A shipment of 7 television sets contains 2 defective sets. A hotel makes a random purchase of 3 of the sets.

If  $x$  is the number of defective sets purchased by the hotel, find the probability distribution of  $X$ .

Find the cumulative distribution function of the random variable  $X$  representing the number of defectives

Then using the cumulative distribution function  $F(X)$ , find

(a)  $P(X = 1)$ ;

(b)  $P(0 < X \leq 2)$ .

## 2.3 Continuous Probability Distributions

### 2.3.1 Probability Density Function (pdf)

- A continuous random variable has a probability of 0 of assuming *exactly* any of its values.
- Consequently, its probability distribution cannot be given in tabular form.
- In this case we are dealing with an interval rather than a point value of our random variable.
- In order to arrive at a probability distribution for a continuous random variable  $X$ , we note that the probability that  $X$  lies between two different values is meaningful.

- Definition 7 - Probability Density Function of a Continuous Random Variable

The function  $f(x)$  (or  $f_X(x)$ ) is a *probability density function* (pdf) of a continuous random variable  $X$ , defined over the set of real numbers  $\mathbb{R}$ , if:

1.  $f(x) \geq 0$  for all  $x \in \mathbb{R}$ ,
2.  $\int_{-\infty}^{\infty} f(x) dx = 1$ ,
3.  $P(a < X < b) = \int_a^b f(x) dx$ .

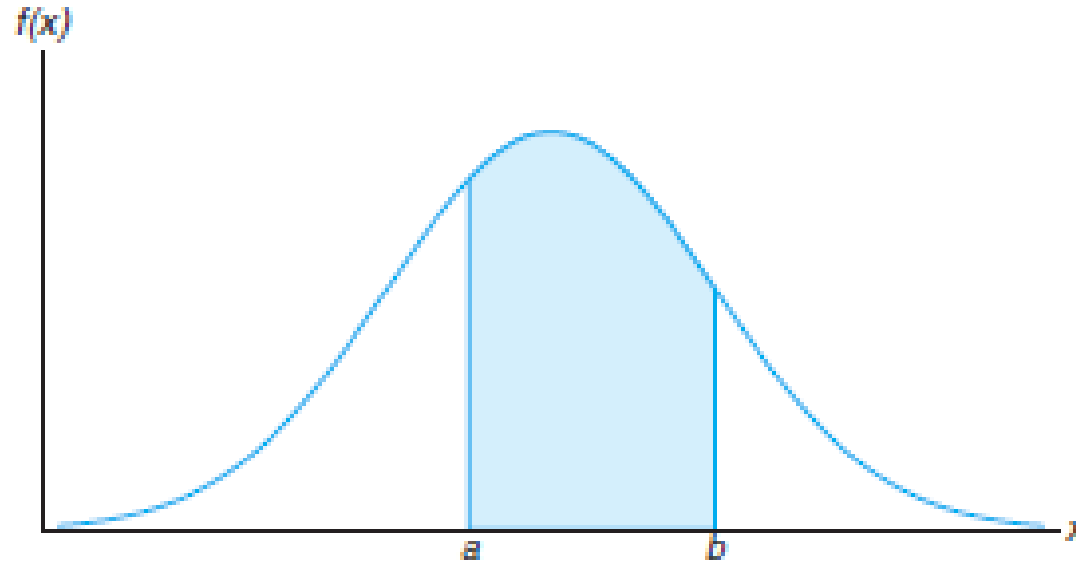
- Also note that, for any continuous random variable  $X$ ,

$$P(a < X \leq b) = P(a < X < b) + P(X = b) = P(a < X < b).$$

That is, it does not matter whether we include an endpoint of the interval or not.

- Note that the probability that  $X$  assumes a value between  $a$  and  $b$  is equal to the shaded area under the probability density function between the ordinates at  $x = a$  and  $x = b$ , and from integral calculus this area is given by

$$P(a < X < b) = \int_a^b f(x) dx.$$



### Example 2.14

- Suppose that the error in the reaction temperature, in  $^{\circ}\text{C}$ , for a controlled laboratory experiment is a continuous random variable  $X$  having the probability density function

$$f(x) = \begin{cases} \frac{x^2}{3}, & -1 < x < 2, \\ 0, & \text{otherwise.} \end{cases}$$

- (a) Verify that  $f(x)$  is a density function.
- (b) Find  $P(0 < X \leq 1)$ .

**Solution:** We use Definition 7.

- (a) Obviously,  $f(x) \geq 0$ . To verify condition 2 in Definition 7, we have

$$\int_{-\infty}^{\infty} f(x) dx = \int_{-1}^2 \frac{x^2}{3} dx = \left[ \frac{x^3}{9} \right]_{-1}^2 = \frac{8}{9} + \frac{1}{9} = 1.$$

(b) Using formula 3 in Definition 7, we obtain

$$P(0 < X < 1) = \int_0^1 \frac{x^2}{3} dx = \left[ \frac{x^3}{9} \right]_0^1 = \frac{1}{9}.$$

### Example 2.15

Find the constant  $c$  such that the function

$$f(x) = \begin{cases} cx^2, & 0 < x < 3 \\ 0, & \text{otherwise} \end{cases}$$

is a probability density function, and compute  $P(1 < X < 2)$ .

**Solution:** Since  $f(x)$  satisfies Property 1 if  $c \geq 0$ , it must satisfy Property 2 in order to be a probability density function.

$$\text{Now, } \int_{-\infty}^{\infty} f(x) dx = \int_0^3 cx^2 dx = \left[ \frac{cx^3}{3} \right]_0^3 = 9c,$$

and since this must be equal to 1 we have  $c = \frac{1}{9}$ .

Using formula 3 in Definition 7, we obtain

$$P(1 < X < 2) = \int_1^2 \frac{1}{9} x^2 dx = \left[ \frac{x^3}{27} \right]_1^2 = \frac{8}{27} - \frac{1}{27} = \frac{7}{27}.$$

Since  $X$  is a continuous random variable, also note that

$$P(1 \leq X \leq 2) = P(1 \leq X < 2) = P(1 < X \leq 2) = P(1 < X < 2) = \frac{7}{27}.$$



### 2.3.2 Cumulative Distribution Function (cdf)

- Definition 8 - Cumulative Distribution Function of a Continuous Random Variable

The *cumulative distribution function*  $F(x)$  (or  $F_X(x)$ ) of a continuous random variable  $X$  with probability density function  $f(x)$  is

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt,$$

where  $x$  is any real number, i.e.,  $-\infty < x < \infty$ .

- As an immediate consequence of Definition 8, one can write the two results

$$P(a < X < b) = F(b) - F(a) \text{ and } f(x) = \frac{dF(x)}{dx},$$

if the derivative exists.

Further note that,  $\lim_{x \rightarrow -\infty} F(x) = 0$ ,  $\lim_{x \rightarrow \infty} F(x) = 1$  and  $0 \leq F(x) \leq 1$ .

### Example 2.16

Again consider the probability density function

$$f(x) = \begin{cases} \frac{x^2}{3}, & -1 < x < 2, \\ 0, & \text{otherwise.} \end{cases}$$

defined in Example 2.14.

Find the cumulative distribution function  $F(x)$  and use it to evaluate  $P(0 < X \leq 1)$ .

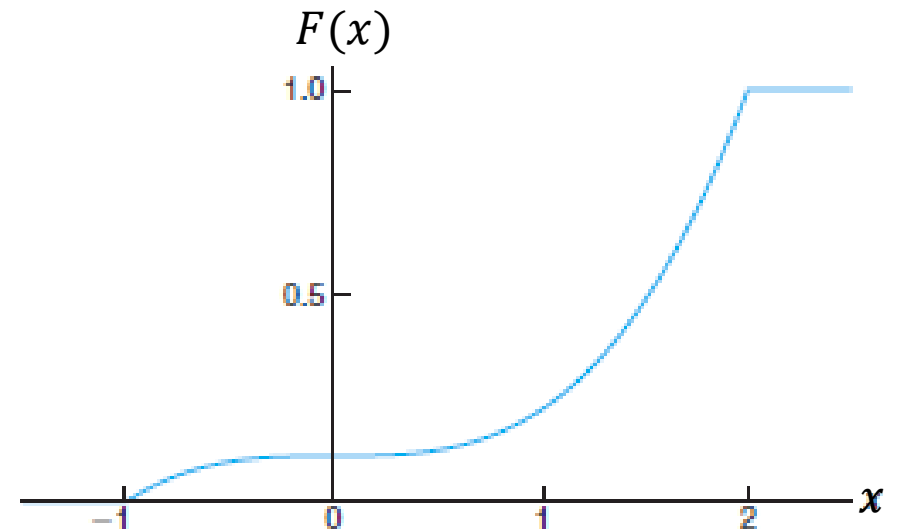
**Solution:** We have  $F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$ .

If  $x < -1$ , then  $F(x) = 0$ .

$$\text{For } -1 < x < 2, \quad F(x) = \int_{-\infty}^x f(t) dt = \int_{-\infty}^x \frac{t^2}{3} dt = \left[ \frac{t^3}{9} \right]_{-1}^x = \frac{x^3 + 1}{9}.$$

$$\begin{aligned}
 \text{If } x \geq 2, \text{ then } F(x) &= \int_{-\infty}^x f(t) dt = \int_{-\infty}^{-1} f(t) dt + \int_{-1}^2 f(t) dt + \int_2^x f(t) dt \\
 &= \int_{-\infty}^{-1} 0 dt + \int_{-1}^2 \frac{t^2}{3} dt + \int_2^x 0 dt \\
 &= \left[ \frac{t^3}{9} \right]_{-1}^2 = 1.
 \end{aligned}$$

$$\text{Therefore, } F(x) = \begin{cases} 0, & x < -1, \\ \frac{x^3+1}{9}, & -1 \leq x < 2, \\ 1, & x \geq 2. \end{cases}$$



The cumulative distribution function  $F(x)$  is given above.

Now,  $P(0 < X \leq 1) = F(1) - F(0) = \frac{2}{9} - \frac{1}{9} = \frac{1}{9}$ , which agrees with the result obtained by using the density function in Example 2.14.

Note that  $F(x)$  increases monotonically from 0 to 1 as is required for a cumulative distribution function.

It should also be noted that  $F(x)$  in this case is continuous.

### Example 2.17

Again consider the probability density function

$$f(x) = \begin{cases} \frac{x^2}{9}, & 0 < x < 3, \\ 0, & \text{otherwise.} \end{cases}$$

defined in Example 2.15.

Find the cumulative distribution function  $F(x)$  for the continuous random variable  $X$  and use it to evaluate  $P(1 < X \leq 2)$ .

**Solution:** We have  $F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$

If  $x < 0$  then  $F(x) = 0$ .

$$\text{For } -1 < x < 3, \quad F(x) = \int_{-\infty}^x f(t) dt = \int_{-\infty}^x \frac{t^2}{9} dt = \left[ \frac{t^3}{27} \right]_{-1}^x = \frac{x^3}{27}.$$

$$\begin{aligned} \text{If } x \geq 3 \text{ then } F(x) &= \int_{-\infty}^x f(t) dt = \int_{-\infty}^0 f(t) dt + \int_0^3 f(t) dt + \int_3^x f(t) dt \\ &= \int_{-\infty}^0 0 dt + \int_0^3 \frac{t^2}{9} dt + \int_3^x 0 dt = \left[ \frac{t^3}{27} \right]_0^3 = 1. \end{aligned}$$

Thus the required cumulative distribution function is

$$F(x) = \begin{cases} 0, & x < 0, \\ \frac{x^3}{27}, & 0 \leq x < 3, \\ 1, & x \geq 3. \end{cases}$$

Note that  $F(x)$  increases monotonically from 0 to 1 as is required for a cumulative distribution function.

It should also be noted that  $F(x)$  in this case is continuous.

We have,  $P(1 < X \leq 2) = F(2) - F(1) = \frac{2^3}{27} - \frac{1^3}{27} = \frac{7}{27}$ , which agrees with the result obtained by using the probability density function in Example 2.15.

**Example 2.18** The Department of Energy (DOE) puts projects out on bid and generally estimates what a reasonable bid should be. Call the estimate  $b$ . The DOE has determined that the probability density function of the winning (low) bid is

$$f(y) = \begin{cases} \frac{5}{8b}, & \frac{2}{5}b \leq y \leq 2b, \\ 0, & \text{elsewhere.} \end{cases}$$

Find cumulative distribution function  $F(y)$  and use it to determine the probability that the winning bid is less than the DOE's preliminary estimate  $b$ .

**Solution:** For  $\frac{2}{5}b \leq y \leq 2b$ , 
$$F(y) = \int_{2/5b}^y \left(\frac{5}{8b}\right) dt = \left[\frac{5t}{8b}\right]_{2/5b}^y = \frac{5y}{8b} - \frac{1}{4}.$$

$$\text{Thus, } F(y) = \begin{cases} 0, & y < \frac{2}{5}b, \\ \frac{5y}{8b} - \frac{1}{4}, & \frac{2}{5}b \leq y < 2b, \\ 1 & y \geq 2b \end{cases}$$

To determine the probability that the winning bid is less than the preliminary bid estimate  $b$ , we have

$$P(Y \leq b) = F(b) = \frac{5}{8} - \frac{1}{4} = \frac{3}{8}.$$



## Exercises

1. The total number of hours, measured in units of 100 hours, that a family runs a vacuum cleaner over a period of one year is a continuous random variable  $X$  that has the probability density function

$$f(x) = \begin{cases} x, & 0 < x < 1, \\ 2 - x, & 1 \leq x < 2, \\ 0, & \text{otherwise.} \end{cases}$$

Find the probability that over a period of one year, a family runs their vacuum cleaner

- (a) less than 120 hours;
- (b) between 50 and 100 hours.

2. The proportion of people who respond to a certain mail-order solicitation is a continuous random variable  $X$  that has the probability density function

$$f(x) = \begin{cases} \frac{2(x+2)}{5}, & 0 < x < 1, \\ 0, & \text{elsewhere.} \end{cases}.$$

- (a) Show that  $P(0 < X < 1) = 1$ .
- (b) Find the probability that more than 1/4 but fewer than 1/2 of the people contacted will respond to this type of solicitation.
3. The waiting time, in hours, between successive speeders spotted by a radar unit is a continuous random variable with cumulative distribution function

$$F(x) = \begin{cases} 0, & x < 0, \\ 1 - e^{-8x}, & x \geq 0. \end{cases}.$$

Find the probability of waiting less than 12 minutes between successive speeders

(a) using the cumulative distribution function of  $X$ ;

(b) using the probability density function of  $X$ .

4. A continuous random variable  $X$  that can assume values between  $x = 1$  and  $x = 3$  has a probability density function given by  $f(x) = \frac{1}{2}$ .

(a) Show that the area under the curve is equal to 1.

(b) Find  $P(2 < X < 2.5)$ .

(c) Find  $P(X \leq 1.6)$ .

Find the cumulative distribution function  $F(x)$  of  $X$ .

5. A continuous random variable  $X$  that can assume values between  $x = 2$  and  $x = 5$  has a probability density function given by  $f(x) = \frac{2(1+x)}{27}$ .

Find (a)  $P(X < 4)$ ;

(b)  $P(3 \leq X < 4)$ .

Find the cumulative distribution function  $F(x)$ , and use it to evaluate  $P(3 \leq X < 4)$ .

6. Consider the probability density function  $f(x) = \begin{cases} k\sqrt{x}, & 0 < x < 1, \\ 0, & \text{otherwise.} \end{cases}$

(a) Evaluate  $k$ .

(b) Find the cumulative distribution function  $F(x)$ , and use it to evaluate  $P(0.3 < X < 0.6)$ .