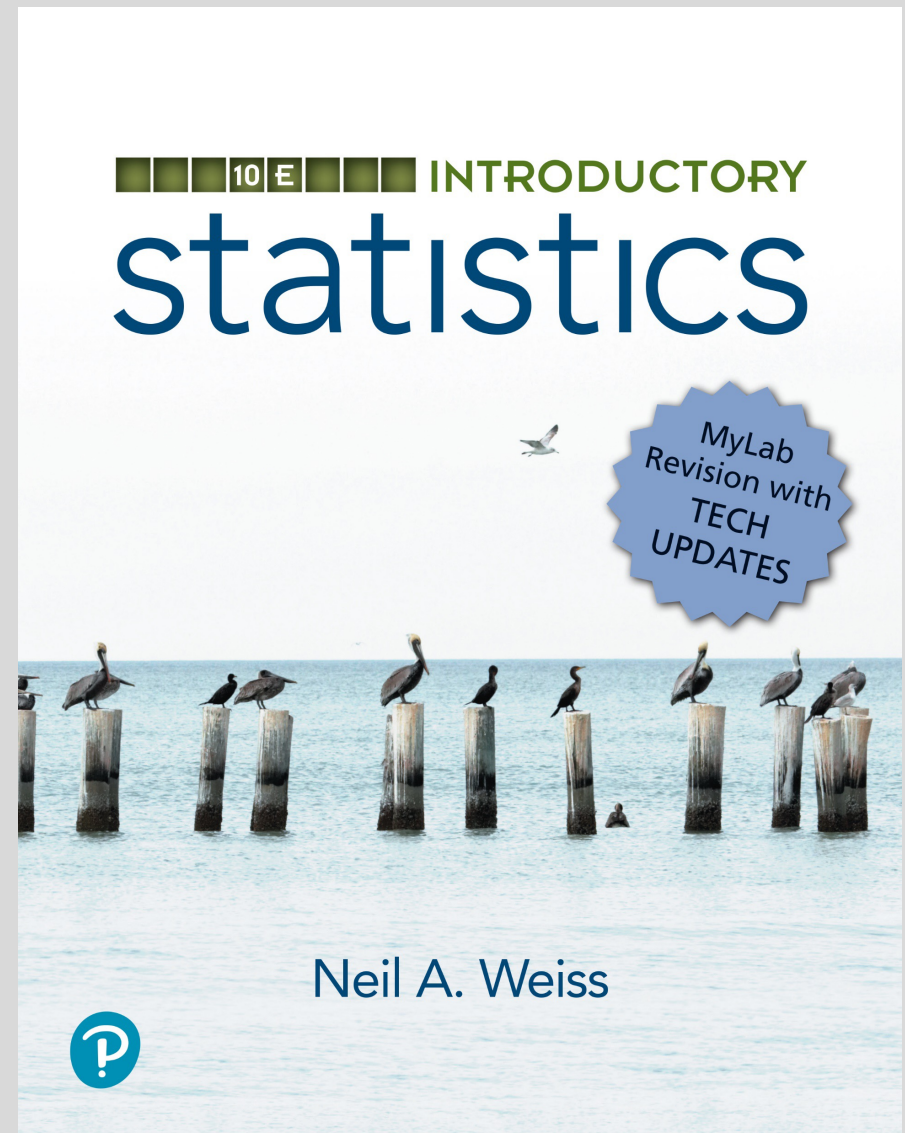


Chapter 5

Discrete Random Variables



Chapter 5

Discrete Random Variables

Section 5.1

Discrete Random Variables and Probability Distributions

Definitions 5.1 & 5.2

Random Variable

A **random variable** is a quantitative variable whose value depends on chance.

Discrete Random Variable

A **discrete random variable** is a random variable whose possible values can be listed. In particular, a random variable with only a finite number of possible values is a discrete random variable.

Definition 5.3

Probability Distribution and Probability Histogram

Probability distribution: A listing of the possible values and corresponding probabilities of a discrete random variable, or a formula for the probabilities.

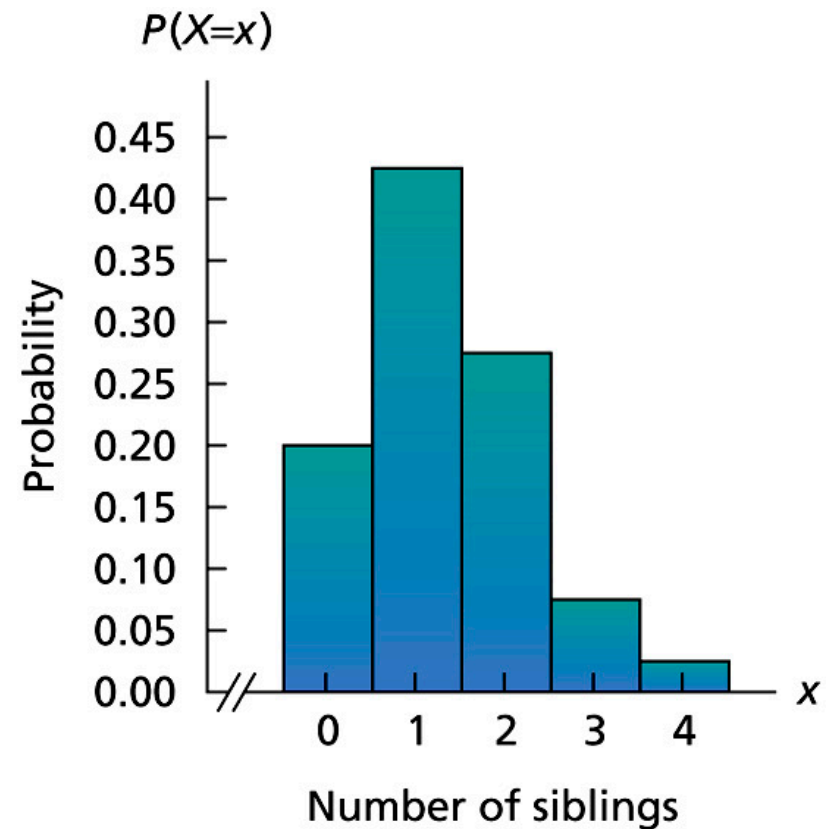
Probability histogram: A graph of the probability distribution that displays the possible values of a discrete random variable on the horizontal axis and the probabilities of those values on the vertical axis. The probability of each value is represented by a vertical bar whose height equals the probability.

Table 5.2 & Figure 5.1

Probability distribution of the random variable X , the number of siblings of a randomly selected student

Siblings x	Probability $P(X = x)$
0	0.200
1	0.425
2	0.275
3	0.075
4	0.025
	1.000

Probability histogram for the random variable X , the number of siblings of a randomly selected student



Key Fact 5.1

Sum of the Probabilities of a Discrete Random Variable

For any discrete random variable X , we have $\sum P(X = x) = 1$.

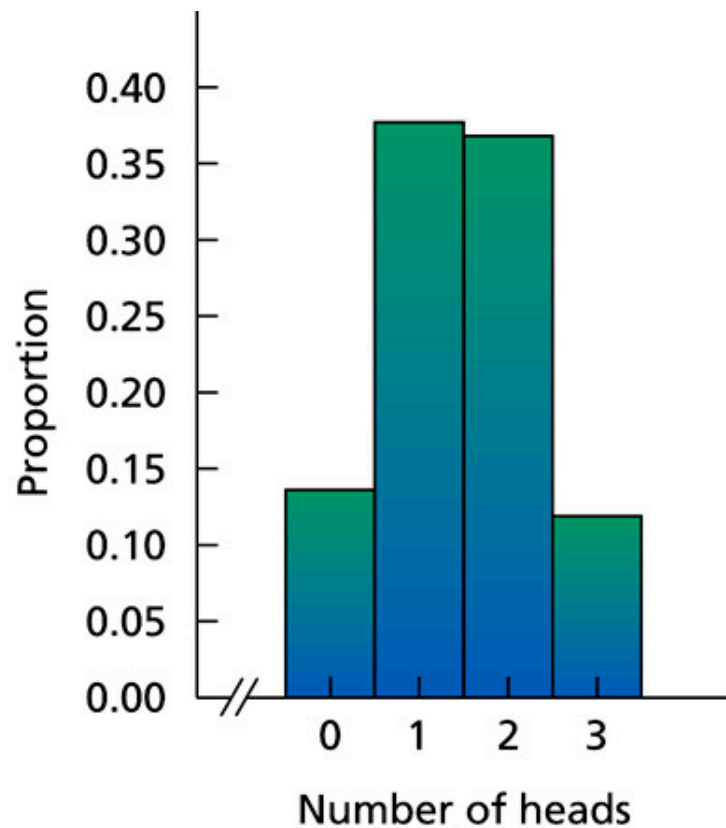
Key Fact 5.2

Interpretation of a Probability Distribution

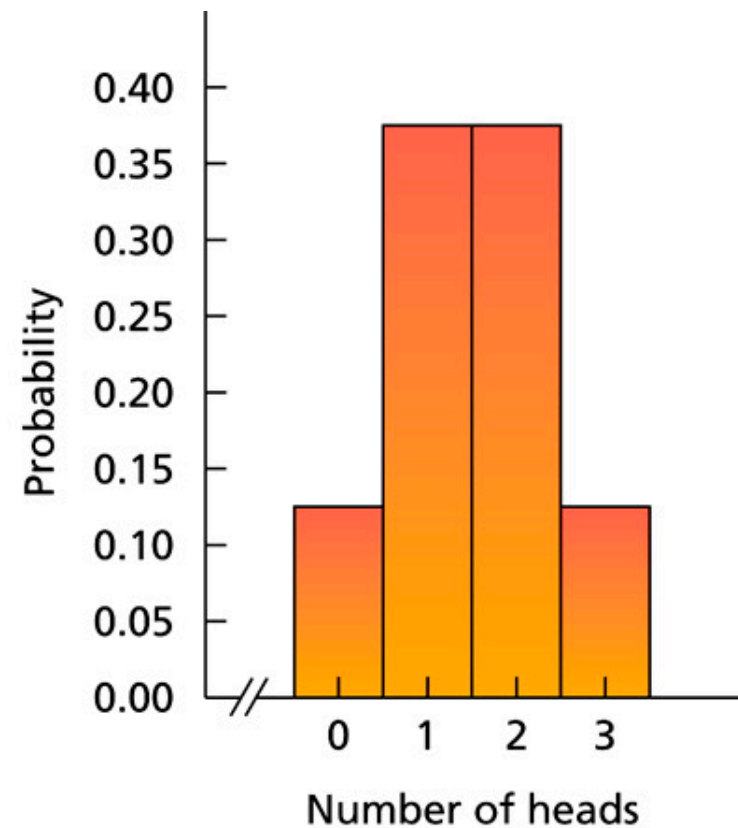
In a large number of independent observations of a random variable X , the proportion of times each possible value occurs will approximate the probability distribution of X ; or, equivalently, the proportion histogram will approximate the probability histogram for X .

Figure 5.2

(a) Histogram of proportions for the numbers of heads obtained in three tosses of a balanced dime for 1000 observations; (b) probability histogram for the number of heads obtained in three tosses of a balanced dime



(a)



(b)

Section 5.2

The Mean and Standard Deviation of a Discrete Random Variable

Definition 5.4

Mean of a Discrete Random Variable

The **mean of a discrete random variable** X is denoted μ_X or, when no confusion will arise, simply μ . It is defined by

$$\mu = \sum x P(X = x).$$

The terms **expected value** and **expectation** are commonly used in place of the term *mean*.[†]

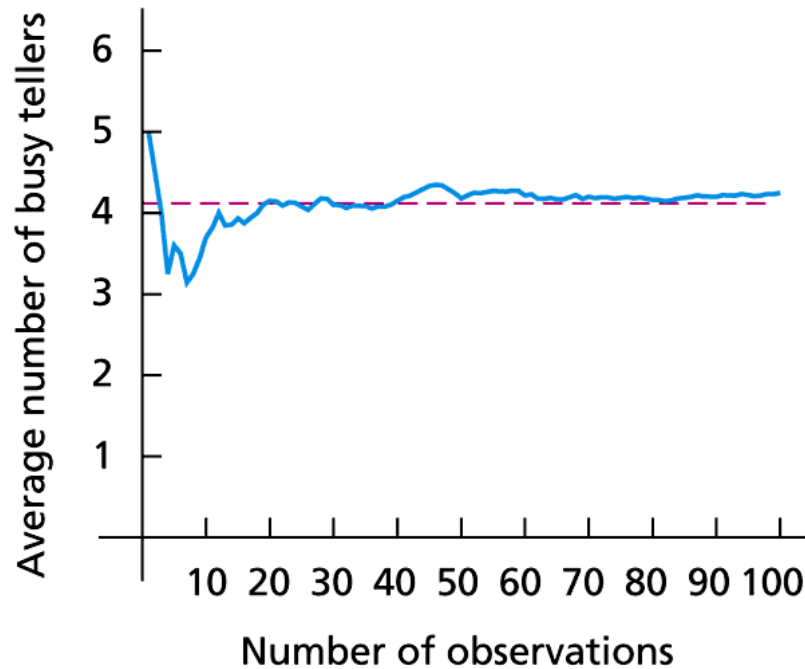
Key Fact 5.3

Interpretation of the Mean of a Random Variable

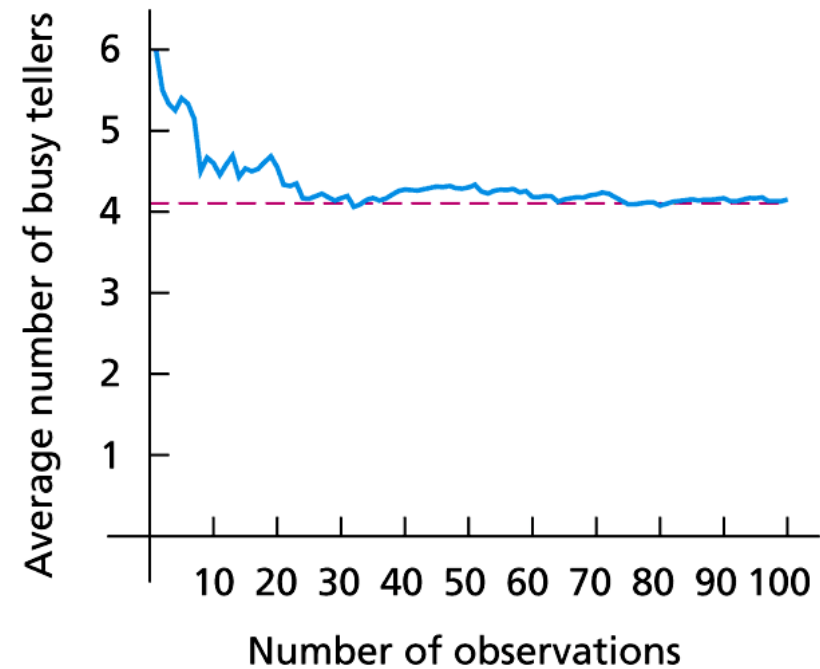
In a large number of independent observations of a random variable X , the average value of those observations will approximately equal the mean, μ , of X . The larger the number of observations, the closer the average tends to be to μ .

Figure 5.3

Graphs showing the average number of busy tellers versus the number of observations for two simulations of 100 observations each



(a)



(b)

Definition 5.5

Standard Deviation of a Discrete Random Variable

The **standard deviation of a discrete random variable X** is denoted σ_X or, when no confusion will arise, simply σ . It is defined as

$$\sigma = \sqrt{\sum (x - \mu)^2 P(X = x)}.$$

The standard deviation of a discrete random variable can also be obtained from the computing formula

$$\sigma = \sqrt{\sum x^2 P(X = x) - \mu^2}.$$

Section 5.3

The Binomial Distribution

Definition 5.6

Factorials

The product of the first k positive integers (counting numbers) is called **k factorial** and is denoted **$k!$** . In symbols,

$$k! = k(k - 1) \cdots 2 \cdot 1.$$

We also define $0! = 1$.

Definition 5.7

Binomial Coefficients

If n is a positive integer and x is a nonnegative integer less than or equal to n , then the **binomial coefficient** $\binom{n}{x}$ is defined as

$$\binom{n}{x} = \frac{n!}{x! (n - x)!}.$$

Definition 5.8

Bernoulli Trials

Repeated trials of an experiment are called **Bernoulli trials** if the following three conditions are satisfied:

1. The experiment (each trial) has two possible outcomes, denoted generically **s** , for **success**, and **f** , for **failure**.
2. The trials are independent, meaning that the outcome on one trial in no way affects the outcome on other trials.
3. The probability of a success, called the **success probability** and denoted **p** , remains the same from trial to trial.

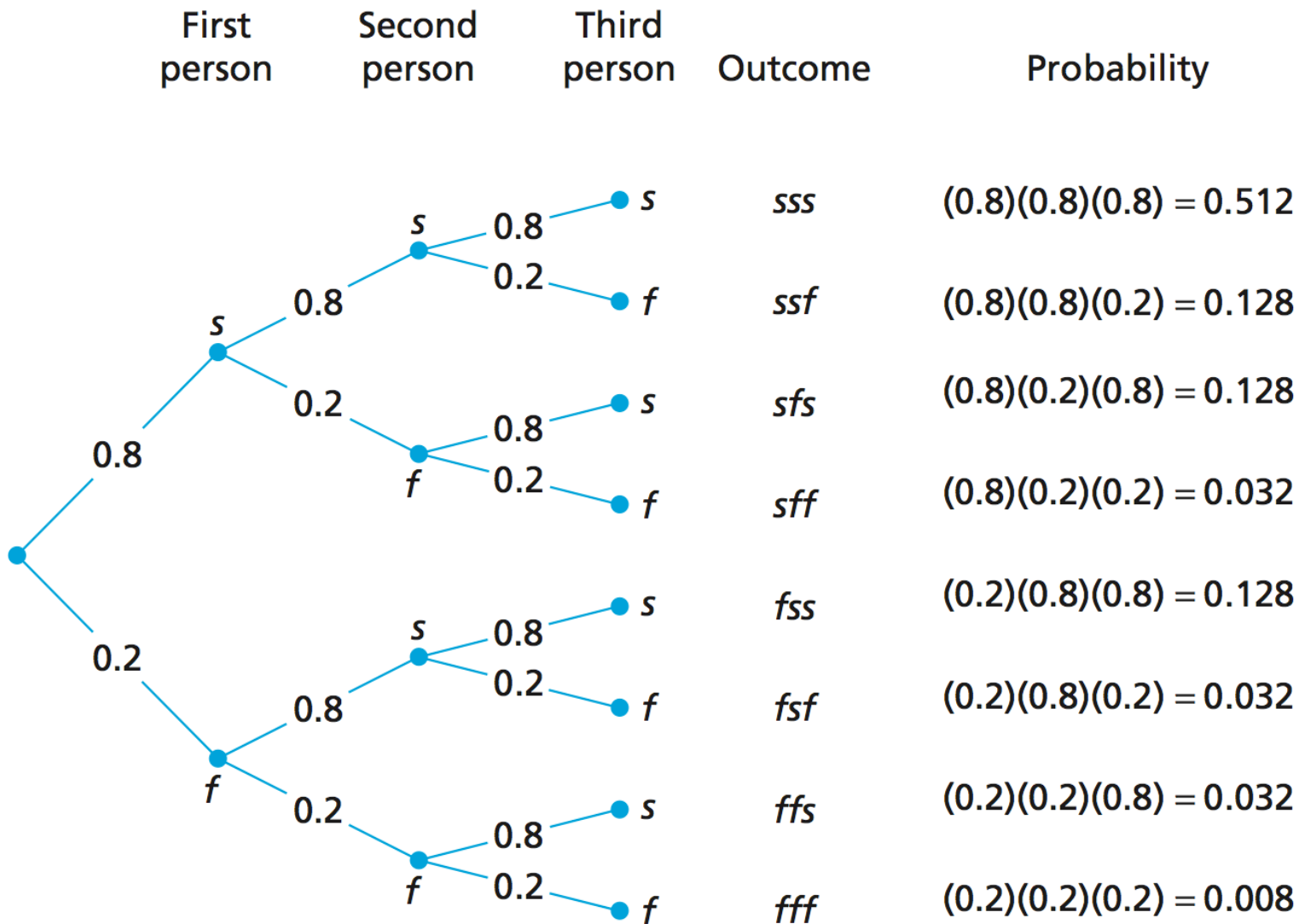
Table 5.14

Outcomes and probabilities for observing whether each of three people is alive at age 65

Outcome	Probability
<i>sss</i>	$(0.8)(0.8)(0.8) = 0.512$
<i>ssf</i>	$(0.8)(0.8)(0.2) = 0.128$
<i>sfs</i>	$(0.8)(0.2)(0.8) = 0.128$
<i>sff</i>	$(0.8)(0.2)(0.2) = 0.032$
<i>fss</i>	$(0.2)(0.8)(0.8) = 0.128$
<i>fsf</i>	$(0.2)(0.8)(0.2) = 0.032$
<i>ffs</i>	$(0.2)(0.2)(0.8) = 0.032$
<i>fff</i>	$(0.2)(0.2)(0.2) = 0.008$

Figure 5.4

Tree diagram corresponding to Table 5.14



Key Fact 5.4

Number of Outcomes Containing a Specified Number of Successes

In n Bernoulli trials, the number of outcomes that contain exactly x successes equals the binomial coefficient $\binom{n}{x}$.

Formula 5.1

Binomial Probability Formula

Let X denote the total number of successes in n Bernoulli trials with success probability p . Then the probability distribution of the random variable X is given by

$$P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x}, \quad x = 0, 1, 2, \dots, n.$$

The random variable X is called a **binomial random variable** and is said to have the **binomial distribution** with parameters n and p .

Procedure 5.1

To Find a Binomial Probability Formula

Assumptions

1. n trials are to be performed.
2. Two outcomes, success or failure, are possible for each trial.
3. The trials are independent.
4. The success probability, p , remains the same from trial to trial.

Step 1 Identify a success.

Step 2 Determine p , the success probability.

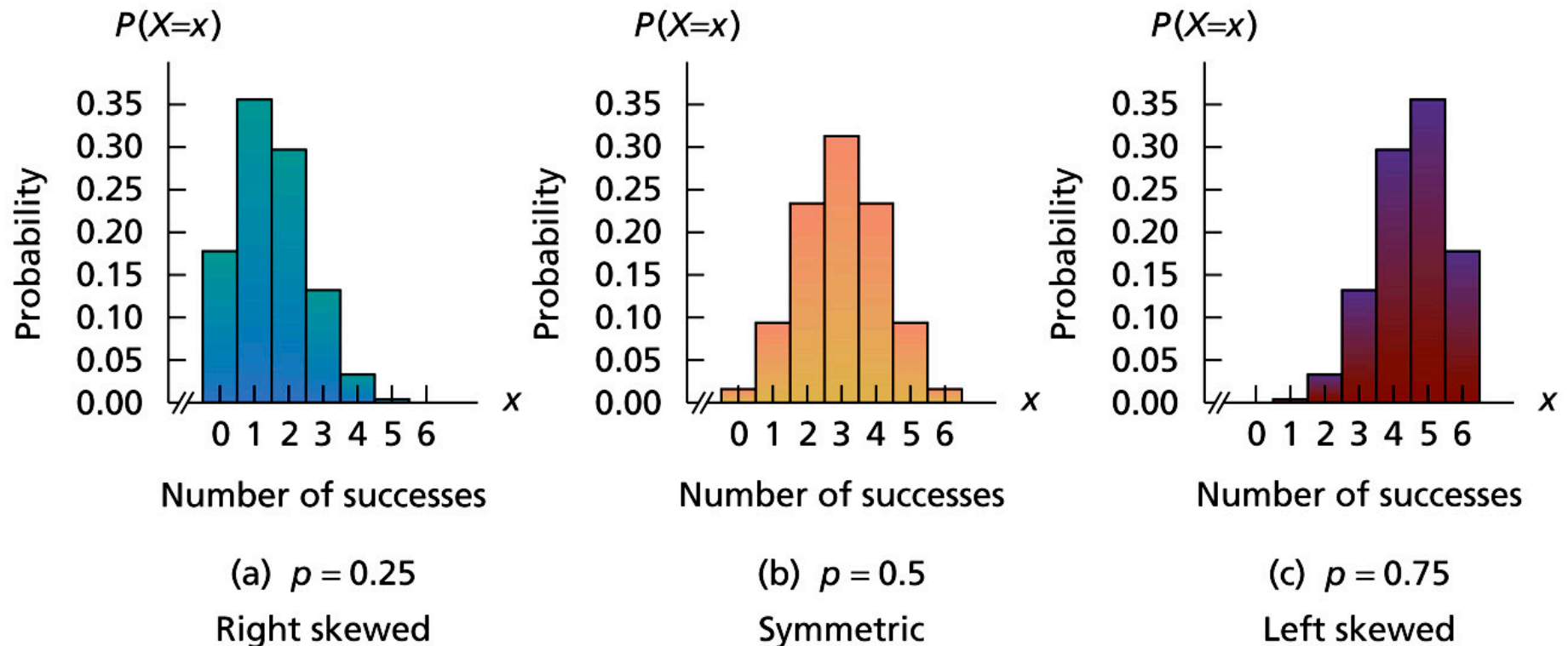
Step 3 Determine n , the number of trials.

Step 4 The binomial probability formula for the number of successes, X , is

$$P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x}.$$

Figure 5.6

Probability histograms for three different binomial distributions with parameter $n = 6$



Formula 5.2

Mean and Standard Deviation of a Binomial Random Variable

The mean and standard deviation of a binomial random variable with parameters n and p are

$$\mu = np \quad \text{and} \quad \sigma = \sqrt{np(1 - p)},$$

respectively.

Key Fact 5.5

Sampling and the Binomial Distribution

Suppose that a simple random sample of size n is taken from a finite population in which the proportion of members that have a specified attribute is p . Then the number of members sampled that have the specified attribute

- has exactly a binomial distribution with parameters n and p if the sampling is done with replacement and
- has approximately a binomial distribution with parameters n and p if the sampling is done without replacement and the sample size does not exceed 5% of the population size.

Section 5.4

The Poisson Distribution

Formula 5.3

Poisson Probability Formula

Probabilities for a random variable X that has a Poisson distribution are given by the formula

$$P(X = x) = e^{-\lambda} \frac{\lambda^x}{x!}, \quad x = 0, 1, 2, \dots,$$

where λ is a positive real number and $e \approx 2.718$. (Most calculators have an e key.) The random variable X is called a **Poisson random variable** and is said to have the **Poisson distribution** with parameter λ .

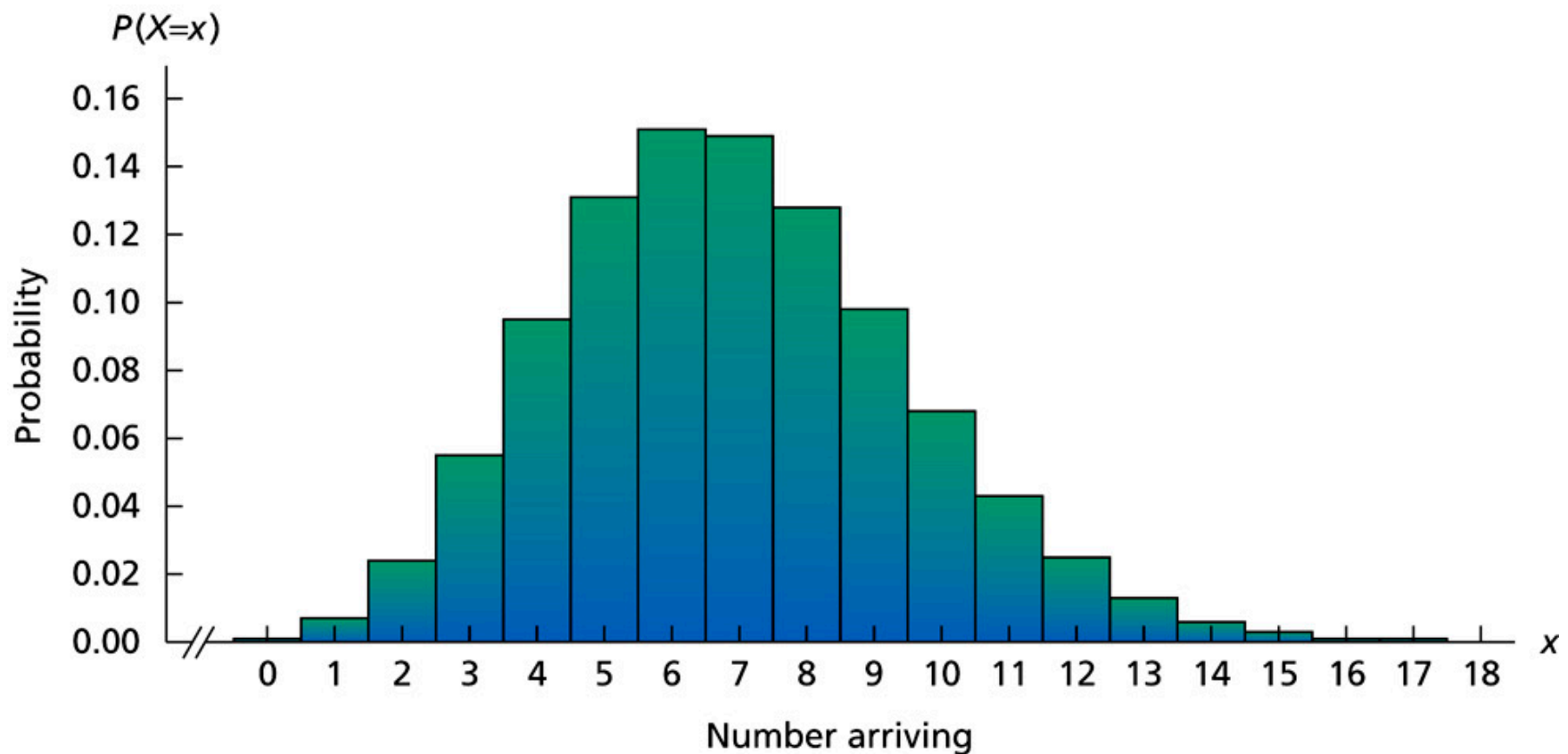
Table 5.16

Partial probability distribution of the random variable X , the number of patients arriving at the emergency room between 6:00 P.M. and 7:00 P.M.

Number arriving x	Probability $P(X = x)$	Number arriving x	Probability $P(X = x)$
0	0.001	10	0.068
1	0.007	11	0.043
2	0.024	12	0.025
3	0.055	13	0.013
4	0.095	14	0.006
5	0.131	15	0.003
6	0.151	16	0.001
7	0.149	17	0.001
8	0.128	18	0.000
9	0.098		

Figure 5.7

Partial probability histogram for the random variable X , the number of patients arriving at the emergency room between 6:00 P.M. and 7:00 P.M.



Formula 5.4

Mean and Standard Deviation of a Poisson Random Variable

The mean and standard deviation of a Poisson random variable with parameter λ are

$$\mu = \lambda \quad \text{and} \quad \sigma = \sqrt{\lambda},$$

respectively.

Procedure 5.2

To Approximate Binomial Probabilities by Using a Poisson Probability Formula

Step 1 Find n , the number of trials, and p , the success probability.

Step 2 Continue only if $n \geq 100$ and $np \leq 10$.

Step 3 Approximate the binomial probabilities by using the Poisson probability formula

$$P(X = x) = e^{-np} \frac{(np)^x}{x!}.$$