Probability Models

TEACHER'S EDITION

DATA-DRIVEN MATHEMATICS

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- The members of the Data-Driven Mathematics leadership team, the consultants, and the writers

- Kathryn Rowe and Wayne Jones for their help in organizing the field-test process and the Leadership Workshops
# Table of Contents

About *Data-Driven Mathematics*  vi
Using This Module  vii

## Unit I: Random Variables and Their Expected Values

<table>
<thead>
<tr>
<th>Lesson</th>
<th>Title</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Probability and Random Variables</td>
<td>3</td>
</tr>
<tr>
<td>2</td>
<td>The Mean as an Expected Value</td>
<td>11</td>
</tr>
<tr>
<td>3</td>
<td>Expected Value of a Function of a Random Variable</td>
<td>18</td>
</tr>
<tr>
<td>4</td>
<td>The Standard Deviation as an Expected Value</td>
<td>24</td>
</tr>
<tr>
<td>Assessment:</td>
<td>Lessons 1–4</td>
<td>32</td>
</tr>
</tbody>
</table>

## Unit II: Sampling Distributions of Means and Proportions

<table>
<thead>
<tr>
<th>Lesson</th>
<th>Title</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>The Distribution of a Sample Mean</td>
<td>39</td>
</tr>
<tr>
<td>6</td>
<td>The Normal Distribution</td>
<td>49</td>
</tr>
<tr>
<td>7</td>
<td>The Distribution of a Sample Proportion</td>
<td>60</td>
</tr>
<tr>
<td>Assessment:</td>
<td>Lessons 5–7</td>
<td>69</td>
</tr>
</tbody>
</table>

## Unit III: Two Useful Distributions

<table>
<thead>
<tr>
<th>Lesson</th>
<th>Title</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>8</td>
<td>The Binomial Distribution</td>
<td>75</td>
</tr>
<tr>
<td>9</td>
<td>The Geometric Distribution</td>
<td>86</td>
</tr>
<tr>
<td>Assessment:</td>
<td>Lessons 8 and 9</td>
<td>95</td>
</tr>
</tbody>
</table>

## Teacher Resources

- **Quizzes** 101
- **Solutions to Quizzes** 108
- **Procedures for Using the TI-83 Graphing Calculator** 113
Historically, the purposes of secondary-school mathematics have been to provide students with opportunities to acquire the mathematical knowledge needed for daily life and effective citizenship, to prepare students for the workforce, and to prepare students for postsecondary education. In order to accomplish these purposes today, students must be able to analyze, interpret, and communicate information from data.

Data-Driven Mathematics is a series of modules meant to complement a mathematics curriculum in the process of reform. The modules offer materials that integrate data analysis with high-school mathematics courses. Using these materials helps teachers motivate, develop, and reinforce concepts taught in current texts. The materials incorporate the major concepts from data analysis to provide realistic situations for the development of mathematical knowledge and realistic opportunities for practice. The extensive use of real data provides opportunities for students to engage in meaningful mathematics. The use of real-world examples increases student motivation and provides opportunities to apply the mathematics taught in secondary school.

The project, funded by the National Science Foundation, included writing and field testing the modules, and holding conferences for teachers to introduce them to the materials and to seek their input on the form and direction of the modules. The modules are the result of a collaboration between statisticians and teachers who have agreed on the statistical concepts most important for students to know and the relationship of these concepts to the secondary mathematics curriculum.

A diagram of the modules and possible relationships to the curriculum is on the back cover of each Teacher's Edition.
Using This Module

Why the Content Is Important

Data analysis is concerned with studying the results of an investigation that has already taken place, with the hope of discovering some patterns in the data that might lead to new insights into the behavior of one or more variables. Probability is concerned with anticipating the future, with the hope of discovering models that might allow the prediction of outcomes not yet seen. Of course, we cannot predict with certainty and so possible outcomes are generally stated along with their chances of occurring.

There is a connection between data and probability since the probabilities used for anticipating future events often come from the analysis of past events. Thus, a survey that says 60% of drivers do not wear seat belts serves as the basis for calculating the probability distribution for the number of drivers, out of the next ten observed, who are not wearing seat belts.

There is also a connection between the key components of describing distributions of data and the key components of describing probability distributions. The mean of the data parallels the expected value of the probability distribution, but notice the change in language from something we see as a fact to something we merely anticipate. Variation in data and in probability distributions is often measured by the standard deviation, but the calculation becomes the expected value of a function of a random variable in the latter case. Shapes of data distributions and probability distributions are described by the same terms—symmetric and skewed: so the context of such descriptions must be made clear.

In this module, students will learn about the connections between data analysis and probability. The emphasis is on the development of basic concepts of probability distributions, as contrasted with probability from counting rules, and the use of standard models for these distributions. The value of having such standard models is that you need study only a few probability distributions in order to solve a wide variety of probability problems. Students will see that a lot of mileage is obtained from the normal, binomial and geometric models.

The skills required for working through this module are mainly those of beginning algebra, except that an infinite series is introduced in Lesson 9. Experience with simulation would be helpful, as many ideas are introduced with this approach.

When teaching this module, it is important to emphasize the distinction between analyzing data to describe the past and building probability models for anticipating the future. We analyze data to discover what happened in the last Gallup poll. We use probability models before the poll is actually conducted to describe what might happen in the next one.
The language of probability requires that a chance event underlie the issue or phenomenon being studied. In this module, chance is thought of in terms of relative frequency. If an unbalanced die has probability 0.4 of coming up a 6, the implication is that after many tosses of the die approximately 40% of the outcomes would be a 6. This may sound obvious, but such a definition of probability rules out the discussion of such issues as "the probability of life on Mars" or "the probability that I will pass this test today."

Probability is sometimes difficult and subtle. It is also useful and fun. If you enter this study in the spirit of data analysis and investigation by simulation, it should be a rewarding educational experience for both you and your students.

**Algebraic Content**
- Variables and random variables
- Functions of random variables
- Summation notation
- Binomial coefficients
- Mathematical models

**Statistics Content**
- Probability distributions
- Expected values
- Standard deviation of a probability distribution
- Normal distribution as a model
- Properties of means and proportions through sampling distributions
- Applications of binomial distribution
- Applications of geometric distribution

**Instructional Model**

*Probability Models* is designed for student involvement and interaction. Each lesson opens with a statement of objectives, followed by some key questions of the type to be addressed in that lesson. The opening scenario is written to foster student discussion and serves to set the stage for the investigation to follow. Investigations almost always involve hands-on activities for the students, and students must work through these to get the full impact of the lesson. Investigations are followed by practice exercises that review and extend the material. Each lesson closes with a summary of the main points.
Working in small groups, students should attempt to answer all of the questions in the discussion sections and work through all of the practice exercises. Each unit is to be completed as a whole, since any one discussion item or problem is likely to build on what went before and may contain new information essential to what follows. This is not a unit in which you can assign the odd numbered exercises; all are important to the understanding of the material.

Although lessons should not be skipped and sections within lessons should not be skipped, it is not necessary to cover the whole module. Unit I covers the basics of random variables and probability distributions, including expected value. Unit II builds on the Unit I material and covers the normal distribution and how it is used as a model for evaluating properties of sample means and proportions. You could stop at this point and still have covered a useful and fairly complete set of lessons on probability distributions as models of reality. Unit III, covering the binomial and geometric distributions, contains lessons that are more specialized and more mathematical.

Whether two or three of the units are covered, the lessons should come close together, as each one in succession tends to build on the one that came before. It is not a good idea to use one of these lessons each Friday on the term, for example. Too much is forgotten in the interim.

**Teacher Resources**

At the back of this Teacher's Edition you will find:

- A quiz for each unit
- Solution keys for the quizzes
- Procedures for Using the TI-83 Calculator

**Where to Use the Module in the Curriculum**

A chapter on probability is usually found somewhere in the algebra sequence, but the material on probability in algebra books is often much abbreviated and weak in modern applications. The two probability modules in *Data-Driven Mathematics*, of which Probability Models is the second, can be used as replacements or supplements for these chapters. Since this material on random variables, probability distributions, and expected values requires subtle reasoning that is not common to most students, it is recommended that this module be used in the second algebra course, or at least no earlier than late in the first algebra course. This is more for the maturity of reasoning required than for any particular set of algebraic skills needed to do the work.
As to requirements, students should be familiar with the idea of a function, especially linear functions of the form $ax + b$, and should have some facility with using symbols. They should also have some experience with basic probability as a relative frequency. Before covering Lesson 8, make sure students have seen binomial expansions.

**Technology**

The calculations required in this module are not as extensive as would be required, for example, in a data analysis module. Nevertheless, students should have at least a graphing calculator available to them. Expected values and related quantities are readily calculated by such a calculator or a computer. Either can also handle the simulations required in this module.
### Pacing/Planning Guide

The table below provides a possible sequence and pacing of the lessons.

<table>
<thead>
<tr>
<th>LESSON</th>
<th>OBJECTIVES</th>
<th>PACING</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Unit I: Random Variables and Their Expected Values</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lesson 1: Probability and Random Variables</td>
<td>Understand the relative frequency concept of probability. Define random variables.</td>
<td>1 class period</td>
</tr>
<tr>
<td>Lesson 2: The Mean as an Expected Value</td>
<td>Understand how to compute and interpret the mean of a probability distribution.</td>
<td>1 class period</td>
</tr>
<tr>
<td>Lesson 3: Expected Value of a Function of a Random Variable</td>
<td>Find expected values of certain functions of random variables. Understand fair games.</td>
<td>1 class period</td>
</tr>
<tr>
<td>Lesson 4: The Standard Deviation as an Expected Value</td>
<td>Understand how to compute and interpret the standard deviation of data and probability distributions.</td>
<td>1–2 class periods</td>
</tr>
<tr>
<td><strong>Unit II: Sampling Distributions of Means and Proportions</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lesson 5: The Distribution of a Sample Mean</td>
<td>Understand the behavior of the distribution of means from random samples.</td>
<td>1 or more class periods depending upon experience with simulation</td>
</tr>
<tr>
<td>Lesson 6: The Normal Distribution</td>
<td>Understand the basic properties of the normal distribution. See the usefulness of the normal distribution as a model for sampling distributions.</td>
<td>1 class period</td>
</tr>
<tr>
<td>Lesson 7: The Distribution of a Sample Proportion</td>
<td>Gain experience working with proportions as summaries of data. Develop sampling distributions for sample proportions. Discover the meaning of margin of error in surveys.</td>
<td>1–2 class periods</td>
</tr>
<tr>
<td><strong>Unit III: Two Useful Distributions</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lesson 8: The Binomial Distribution</td>
<td>Understand the basic properties of the binomial distribution. Use the binomial distribution as a model for certain types of counts.</td>
<td>2 class periods</td>
</tr>
<tr>
<td>Lesson 9: The Geometric Distribution</td>
<td>Understand the basic properties of the geometric distribution. Use the geometric distribution as a model for certain types of counts.</td>
<td>1–2 class periods</td>
</tr>
</tbody>
</table>
Random Variables and Their Expected Values
LESSON 1

Probability and Random Variables

Materials: none
Technology: graphing calculators or computers (optional)
Pacing: 1 class period with extra time for homework

Overview

The lesson begins by looking at the distribution of the number of children per family, which is a variable taking on a finite number of integer values—discrete variables. The relative frequencies of the possible values become the estimated probabilities for those values as the lesson moves from data description to probability models. This connection, along with the distinction between the relative frequencies of data and the probabilities associated with a random variable, are the key parts of the lesson. The addition of probabilities for mutually exclusive events and complements are two basic probability concepts that are used with little introduction or explanation. These concepts may have to be reviewed.

Teaching Notes

Have students work through the investigations and practice exercises in small groups, if possible, with little instruction. In the discussions, make sure students understand the concepts of random variable—as distinguished from an ordinary variable taking on numerical values—and probability distribution. Also be sure that they understand what is meant by a model.

Follow-Up

Have students look for other distributions of data for variables of this discrete type. Have them display and plot the data, and describe a scenario in which the data distribution could serve as an estimate of a probability distribution.
LESSON 1: PROBABILITY AND RANDOM VARIABLES

STUDENT PAGE 3

Probability and Random Variables

How many children are in a typical American family?

What is the probability of randomly choosing a family with two children?

What is a random variable?

According to the U.S. Bureau of the Census, the number of children under 18 years of age per family has a distribution as given on the table below. A "family" is defined as a group of two or more persons related by birth, marriage, or adoption, residing together in a household. In which category does your family belong?

INVESTIGATE

Family Size

In reality, some families have more than four children under the age of 18. However, the number of such families is very small and their percent would be very small compared to the percents in this table. Thus, we can describe the essential features of the number of children per family by using this simplified table as a model of reality.

<table>
<thead>
<tr>
<th>Number of Children</th>
<th>Percent of Families</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>51</td>
</tr>
<tr>
<td>1</td>
<td>19</td>
</tr>
<tr>
<td>2</td>
<td>19</td>
</tr>
<tr>
<td>3</td>
<td>7</td>
</tr>
<tr>
<td>4</td>
<td>3</td>
</tr>
</tbody>
</table>
Solution Key

Discussion and Practice

1. a. 20%
   
   b. $100 - 51 = 49$
   
   c. Seven percent of families in the U.S. have exactly 3 children under the age of 18.
   
   d. The percents add to 100, which is appropriate if all categories of children per family are accounted for. The paragraph that follows explains that this is an approximation.

2. a. The height of the bar is about 19 units, which represents the 19% of families that have exactly 2 children under the age of 18.
   
   b. $51 + 20 + 19 = 90$
   
   c. The distribution of number of children per family has a high point at 0 and then decreases through 4. The percent of families with more than 4 children is so small as to not show up on the table or graph. The distribution can be called skewed toward the greater numbers of children, that is, to the right.

Discussion and Practice

The first line of data in the table above is interpreted to mean that 51% of the families in the United States have no children under the age of 18.

a. What percent of families have one child under the age of 18?

b. What percent of families have at least one child under the age of 18?

c. How would you interpret the 7% for the “3 children” category?

d. What is the sum of the percents in the table? Explain why this is an appropriate value for the sum.

2. A bar graph of the data on the number of children per family is shown below.

   a. What does the height of the bar over the 2 represent?
   
   b. What percent of families have at most two children under the age of 18?
   
   e. Describe in words the distribution of children per family.

Random Variables

The discussion above makes use of the data table to describe one aspect of families in the United States. Suppose A. C. Nielsen, the company that provides ratings of TV shows, is planning to select a random sample of families from across the country. In that case, these same percents can be used as probabilities so that Nielsen can anticipate how many children under the age of 18 they might encounter in the sample.
3. a. 0.20
   b. 1 - 0.51 = 0.49
   c. 0.51 + 0.20 + 0.19 = 0.90
   d. 0.19 + 0.07 = 0.26
   e. Based on the information given, the probability that a family has 5 children is 0.00. This does not imply that a family cannot have 5 children under the age of 18. The probability of such an event occurring could be, say, 0.002, which rounds to 0.00. Remember that the data in the table are only approximations to the true probabilities.

4. a. \( P(C = 1) \)
   b. \( P(C \geq 1) \)
   c. \( P(C \leq 2) \)
   d. \( P(C = 2) \) or \( (C = 3) \) = \( P(C = 2) \) + \( P(C = 3) \)
   e. \( P(C = 5) \)

5. a. 0.51
   b. 0.51 + 0.20 + 0.19 = 0.90
   c. 0.07 + 0.03 = 0.10 = 1 - 0.90
   d. 0.20 + 0.19 + 0.07 = 0.46

6. a. The probability that a randomly selected family has either 1 or 3 children under the age of 18
b. The probability that a randomly selected family has 2 or more children under the age of 18

### Student Page 5

3. Suppose Nielsen is to select one family at random. What is the approximate probability that the selected family will have:
   a. exactly one child under the age of 18?
   b. at least one child under the age of 18?
   c. at most two children under the age of 18?
   d. either two or three children under the age of 18?
   e. exactly five children under the age of 18?

In your past work, symbols have helped you to communicate mathematical statements more clearly and more concisely. Symbols can also help to clarify probability statements. In the situation above, the numerical outcome of interest is "the number of children under the age of 18 in a randomly selected U.S. family." Instead of writing this long statement each time we need it, why not just call it \( C \)? Then, \( C \) = the number of children under the age of 18 in a randomly selected U.S. family. From the data table, you can see that the probability that \( C \) is equal to 1 is 0.20, or 20%. It is cumbersome to write this probability statement in words, so we use a shorthand notation for the statement. The symbolic statement is \( P(C = 1) = 0.20 \).

When probability statements involve intervals of values for \( C \), the symbolic form makes use of inequalities. For example, the probability that a randomly selected family has "at most one child under the age of 18" implies that the family has "either 0 or 1 child under the age of 18." This can be written as
\[
P(C \leq 1) = P(C = 0) + P(C = 1) = 0.51 + 0.20 = 0.71
\]

4. Write a symbolic statement for each of the statements in Problem 3.
   a. \( P(C = 0) \)
   b. \( P(C \leq 2) \) = \( P(C = 0) \) or \( (C = 1) \) or \( (C = 2) \)
   c. \( P(C \geq 3) \)
   d. \( P(1 \leq C \leq 3) \)

5. Write each of the following symbolic statements in words.
   a. \( P(C = 1) \) or \( (C = 3) \)
   b. \( P(C \geq 2) \)
c. The probability that a randomly selected family has between 2 and 4 (2, 3, or 4) children under the age of 18.
d. The probability that a randomly selected family has either 2 or fewer or 4 or more children under the age of 18 or The probability that a randomly selected family does not have exactly 3 children under the age of 18.

7. a. \( P(C = 0 \text{ or } C = 2 \text{ or } C = 4) \) = 0.51 + 0.19 + 0.03 = 0.73
b. \( P(C < 2) = P(C \leq 1) = 0.51 + 0.20 = 0.71 \)
c. \( P(C \leq 1) = 0.51 + 0.20 = 0.71 \)
d. \( P(C = 3) = 0.07 \)
8. \( P(C \leq 3) = 1 - 0.03 = 0.97 \)

The symbols like \( C \) used to represent numerical outcomes from chance processes are called random variables. Random variables are the basic building blocks for working with probability in scientific investigations. Probability distributions for random variables can be conveniently displayed in a two-column table like the one shown below for the random variable \( C \), the "number of children under 18 in a randomly selected family."

<table>
<thead>
<tr>
<th>( C )</th>
<th>( P(C) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.51</td>
</tr>
<tr>
<td>1</td>
<td>0.20</td>
</tr>
<tr>
<td>2</td>
<td>0.19</td>
</tr>
<tr>
<td>3</td>
<td>0.07</td>
</tr>
<tr>
<td>4</td>
<td>0.03</td>
</tr>
</tbody>
</table>

The probability distribution for a random variable can also be displayed in a bar graph, like the one shown below for the random variable \( C \).
9. a. Possible answer: The probability distribution of number of children per family has maximum at 0 and then decreases through 4. The probability of a randomly selected family having more than 4 children is so small as to not show up on the table or graph. The probability distribution can be called skewed toward the greater numbers of children, that is, skewed to the right.

b. Possible answer: The figures in Problems 2 and 8 have the same shape. In Problem 2, the vertical scale is percent, while in Problem 8 it is probability, which is measured on the interval 0 to 1. The distribution in Problem 2 describes the Census data on number of children per family. The distribution in Problem 8 describes the chances of getting a family with a certain number of children by random selection from among the families in the U.S.

c. The probabilities add to 1, as they account for a complete set of non-overlapping outcomes to the event “Select a family at random from the population of the U.S.” It is possible to select a family with more than 5 children, but the probability would round to 0.00.

Practice and Applications

10. a. No. Possible answer: A household could have more than 4 cars, but the percent of such households would round to 0.0.
LESSON 1: PROBABILITY AND RANDOM VARIABLES

STUDENT PAGE 8

An automobile manufacturer is planning to conduct a survey on what Americans think about automobile repairs. What is the chance that a randomly selected household in the poll has

a. no cars?
b. exactly one car?
c. at least one car?
d. at most two cars?
e. exactly three cars?

11. Write each of the statements in Problem 11 in symbolic form.

12. With Y defined as in Problem 10, find each of the following probabilities and write the symbolic statement in words.
a. P(Y = 2)
b. P(Y ≥ 0)
c. P(Y ≤ 3)
d. P(Y = 0)
e. P(Y = 1)

13. a. 0.437; what is the probability that a randomly selected household has exactly two cars?
b. 1.0; what is the probability that a randomly selected household either has a car or does not have a car?
c. 1 - 0.106 = 0.894; what is the probability that a randomly selected household has at most three cars?
d. 0.228 + 0.437 + 0.215 = 0.880; what is the probability that a randomly selected household has one, two, or three cars?

14. P(Y ≥ 1) = 1 - P(Y = 0) = 1 - 0.014 = 0.986
15. **a.** Yes, for each age group, the percent of men who had never married is greater than the percent of women who had never married. This is true for both 1970 and 1991.

**b.** \(1 - 0.64 = 0.36\)

**c.** 0.47

**d.** The data give only the percent never married among the males or females in a certain age class. No information is provided on the number of people in the various age classes. Therefore, we cannot estimate the probability that a randomly selected adult would be under age 34.

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**SUMMARY**

A display, such as a table or a graph, showing the numerical values that a variable can take on and the percent of time that the variable takes on each value is called the distribution of that variable. If possible values of the variable are randomly selected, the variable is called a random variable and the percents attached to the numerical values give the probability distribution for that variable.
The Mean as an Expected Value

Materials: none
Technology: graphing calculators (optional)
Pacing: 1 class period

Overview
The mean of a set of data arranged in a frequency table can be calculated by knowing either the frequencies or the relative frequencies for each data value. The relative frequencies may be displayed as a percent or as a fraction or decimal. When the relative frequencies are viewed as probabilities of future events, the calculation for the mean remains the same but the result is called the expected value of the random variable. A general formula for the expected value is developed using summation notation.

Teaching Notes
Allow students to work through the lesson in small groups. The summation notation used in the general formula for expected value will require some discussion, especially if it has not been seen before. Make sure students understand the subtleties of the difference between a mean of a data distribution and the expected value of a probability distribution.

Technology
Here, a graphing calculator is extremely useful for calculating expected values. For example, the TI-83 allows you to calculate the expected value in one step if the data values are in one list and the relative frequencies in another, using decimals. Use the STAT/CALC/1-Var Stats command with the data list named first and the relative frequency list second.

Follow-Up
Have students suggest other data sets or probability distributions for which the expected value would be a meaningful summary.
The Mean as an Expected Value

What is the average number of children per family in America?

In a randomly chosen family, how many children would you expect to see?

How does the mean number of children per family compare to the mean family size?

An average, such as the arithmetic mean or simply the mean, is a common measure of the center of a set of data. The mean score of your quizzes in mathematics is, no doubt, an important part of your grade in the course. The mean age of residences in your neighborhood helps insurance companies figure out how much to charge for fire insurance. The mean amount paid by families for typical goods and services this year as compared to last year determines the rate of inflation. In this lesson, we will look at means of distributions of data to discover how they relate to means of probability distributions.

INVESTIGATE
How would you calculate the mean score of your quizzes in mathematics? In what other situations might you need to calculate the mean for a set of data?

Discussion and Practice
1. A football team played nine games this season, scoring 12 points in each of three games and 21 points in each of the other six games.
LESSON 2: THE MEAN AS AN EXPECTED VALUE

Solution Key
Discussion and Practice

1. a.

![Bar graph]

Number of Games

<table>
<thead>
<tr>
<th>Scores</th>
<th>Number of Games</th>
</tr>
</thead>
<tbody>
<tr>
<td>12</td>
<td>2</td>
</tr>
<tr>
<td>21</td>
<td>6</td>
</tr>
</tbody>
</table>

b. 18

c. The mean is closer to 21 since this score has the higher frequency and, hence, draws the mean toward it.

2. a.

![Bar graph]

Possible answer: It is similar except that the vertical axis represents fraction of games played rather than number of games played.

b. \(12\left(\frac{1}{3}\right) + 21\left(\frac{2}{3}\right) = 18\)

c. Again, the mean is closer to the score of 21, which has the greater relative frequency.

3. The expected score for the future is the mean score from the past, 18.

4. Possible answer: The actual data are available for this year and the true mean score can be calculated. This year's record is used as a probability distribution for next year's record. In this sense, it tells us what is anticipated, or expected, to happen. The expected value is the mean score we anticipate, or expect, at this point, but it may not actually occur.

5. Yes; the expected number of points scored against the best team should be a little less than the average number of points scored against all opponents.

Expected Value
Recall that one of the first numerical summaries of a set of data that you studied was the mean, used as a measure of center. We now review the calculation of the mean by working through an example. A survey of a class of 20 students reveals that 4 have no pets, 10 have one pet, and 6 have two pets. The data are shown in the table below.

a. Construct a bar graph for the points scored, with the values for the variable "points scored" on the horizontal axis.

b. What is the mean number of points scored per game for this team? Explain how you found this mean.

c. Mark the value of the mean on the horizontal axis of the bar graph. Is the mean closer to 12 or to 21?

d. Suppose you knew that the team scored 12 points in \(\frac{1}{3}\) of its games and 21 points in \(\frac{2}{3}\) of its games, but you were not told how many games the team played.

a. Construct a bar graph for these data. How does it compare to the one in Problem 1a?

b. Can you still calculate the mean number of points per game? If so, what is it? Discuss how you arrived at this answer.

c. Mark the mean on the horizontal axis of the bar graph.

d. The team is expected to perform next year about as well as it performed this year. That is, the probability of scoring 12 points in a game is about \(\frac{1}{3}\), while the probability of scoring 21 points in a game is about \(\frac{2}{3}\). For a randomly selected game, how many points would you expect the team to score?

a. A mean computed from a probability distribution—an anticipated distribution of outcomes—is called an expected value. Discuss why you think this terminology is used. Does the terminology seem appropriate?

b. Instead of a randomly selected game from next year's schedule, suppose we consider the game against the best team in the league. Would that change your opinion on the team's expected number of points scored? Why or why not?
6. \( \frac{22}{20} = 1.1; \) 0(\( \frac{4}{20} \)) + 1(\( \frac{10}{20} \)) + 2(\( \frac{6}{20} \)) = 1.1

7. a. 20%, 50%, 30%
   b. 0(0.2) + 1(0.5) + 2(0.3) = 1.1
   c. The answers are the same, based on either frequencies or relative frequencies.

8. a. 0.91
   b. The mean is not in the center of the graph. It is much closer to 0 because of the high frequency there.

The mean number of pets per student can be calculated in a variety of ways.

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The mean number of pets per student can be calculated in a variety of ways.
Lesson 2: The Mean as an Expected Value

9. \( E(C) = 0.91 \)

10. Problem 8 asks about the actual mean number of children in the current population of families. Problem 9 asks about the anticipated mean number of children in a random sample of families yet to be selected.

11. a. The expected value is the average value of the number of children per family that we anticipate seeing after many families are selected in a sample. The average number of children per family, across many families, need not be an integer.

   b. \( 100(0.91) = 91 \)
   c. \( 2500(0.91) = 2275 \)
   d. \( 4000 = 0.91n; n = 4000/0.91 = 4396 \)

12. Suppose a random variable \( X \) can take on values \( x_1, x_2, ..., x_k \) with respective probabilities \( p_1, p_2, ..., p_k \). That is, \( P(X = x_i) = p_i \) for values of \( i \) ranging from 1 to \( k \).
12. a. $E(X) = x_1p_1 + x_2p_2 + \ldots + x_kp_k$
   The mean is found by multiplying each value of $x$ by its relative frequency and adding the results. In a probability distribution, the expected value is found by the same procedure with the probabilities replacing the relative frequencies, since probability is a long-run relative frequency.
   b. $E(X) = \sum_{i=1}^{k} x_i(p_i)$

**Practice and Applications**

13. a. $E(Y) = 2.58$, much greater than the expected number of children per family
   b. $1000(2.58) = 2580$
   c. $n = 4000/2.58 \approx 1550$

---

14. a. Write a symbolic expression for the expected value of $X$. Explain the reasoning behind this expression.
   b. A commonly used symbol for the expected value of $X$ is $E(X)$, and $E(X)$ is expressed as a sum. The $\Sigma$ symbol tells you to add the terms that follow the symbol, starting with the term indicated by the integer below the $\Sigma$ and ending with the term indicated by the integer above the $\Sigma$. Thus,
   $$\sum_{i=1}^{k} x_i(p_i)$$
   Replace the question marks in the expression below with numerical values to indicate the range of summation.
   $$f(X) = \sum_{i=1}^{k} x_i$$

**Practice and Applications**

The following table shows the distribution of household sizes for U.S. households.

<table>
<thead>
<tr>
<th>Number of Persons per Household</th>
<th>Percent of Households</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>25</td>
</tr>
<tr>
<td>2</td>
<td>22</td>
</tr>
<tr>
<td>3</td>
<td>17</td>
</tr>
<tr>
<td>4</td>
<td>16</td>
</tr>
<tr>
<td>5</td>
<td>7</td>
</tr>
<tr>
<td>6</td>
<td>2</td>
</tr>
<tr>
<td>7</td>
<td>1</td>
</tr>
</tbody>
</table>

Note: Households of more than 7 persons are very rare

15. Suppose Nielsen is randomly sampling households in order to produce TV ratings. Let $Y$ denote the size of a randomly selected household.
   a. Find the expected value of $Y$. Compare it to the expected value of $C$ found in Problem 9.
   b. If Nielsen randomly selects 1000 households, how many people would these households be expected to contain?
   c. If Nielsen really expects 4000 people in the survey, how many households should be sampled?
The expected value of 2.58 is to the left of center on the graph because the high probabilities at 1 and 2 pull the expected value in their direction.

14. Construct a bar graph of the probability distribution for Y. Mark the expected value of Y on the bar graph. Is the expected value in the center of the possible values for Y? Why or why not?

**SUMMARY**

The distribution of a variable is determined by the numerical values that the variable can take on, along with the proportion of times that each numerical value occurs. The mean of the distribution can be calculated from this information. If the proportions of times that the numerical values occur are interpreted as probabilities, then the mean is called the expected value of a probability distribution. The mean of a data set describes something that has already happened, while an expected value anticipates what might happen in the future.
LESSON 3

Expected Value as a Function of a Random Variable

Materials: none
Technology: scientific or graphing calculators (optional)
Pacing: 1 class period

Overview
The two goals of this lesson are to demonstrate that the expected value of a linear function is the same linear function of the expected value and to understand what is meant by a fair game. Payoffs for games often can be written as linear functions of random variables, and so the two topics naturally fit together.

Teaching Notes
As in previous lessons, these exercises are appropriate as small-group activities. You should direct some class discussion around the issue of fair games and how this idea applies to business decisions, for example.

Follow-Up
Have students construct some games of their own, perhaps using spinners or dice, for which they can calculate the probability of winning. Have them decide whether or not the games are fair.
LESSON 3: EXPECTED VALUE AS A FUNCTION OF A RANDOM VARIABLE

STUDENT PAGE 16

Expected Value of a Function of a Random Variable

How much would you expect to pay to feed a pet for a week?

What is a “fair” game?

How do business decisions depend on the concept of a fair game?

OBJECTIVES
Find expected values of certain functions of random variables.
Understand fair games.

As you have seen in previous work in mathematics and science, it is often convenient to express one variable as a function of another. Most goals in basketball are worth 2 points; hence, the number of points a player scores in a game, excluding free throws and 3-point goals, can be written as a function of the number of goals made.

Charlotte takes about 20 shots from inside the 2-point area during a game and expects to make about 60% of them. How many points can she expect to get from these goals?

INVESTIGATE
Expected Value of a Function

In practical applications of probability, the random variables are often written as functions of other random variables.

Suppose the table below gives estimates of the probabilities of a randomly selected student having 0, 1, or 2 pets.
**Solution Key**

**Discussion and Practice**

1. \( E(X) = 0(0.2) + 1(0.5) + 2(0.3) 
   = 1.1 \)

2. a.
   \[ \begin{array}{c|c}
   Y & P(Y) \\
   \hline
   0 & 0.2 \\
   20 & 0.5 \\
   40 & 0.3 \\
   \end{array} \]

   b. \( E(Y) = 0(0.2) + 20(0.5) + 40(0.3) = 22 \)

3. a. \( E(Y) = \sum_{i=1}^{k} y_i(p_i) = \sum_{i=1}^{k} 20x_i(p_i) \)
   
   \[ = 20 \sum_{i=1}^{k} x_i(p_i) = 20E(X) \text{ since } \]
   
   the values for \( Y \) are simply 20 times the corresponding values for \( X \).

   b. \( E(Y) = 20E(X) = 20(1.1) = 22 \)

4. \( E(\text{Profit}) = E(\text{Gain}) - 20 = 15(8) - 20 = 100 \)

5. a. \( E(Y) = \sum_{i=1}^{k} y_i(p_i) \)

   b. \[ \sum_{i=1}^{k} y_i(p_i) = \sum_{i=1}^{k} (ax_i + b)(p_i) \]

   c. \[ \sum_{i=1}^{k} (ax_i + b)(p_i) \]
   
   \[ = a \sum_{i=1}^{k} x_i(p_i) + b \sum_{i=1}^{k} (p_i) \]

   d. \[ a \sum_{i=1}^{k} x_i(p_i) + b \sum_{i=1}^{k} (p_i) \]
   
   \[ = aE(X) + b \]

   Note that the sum of the probabilities must be 1.

If \( X \) represents the number of pets per randomly selected student, then \( E(X) \) can be calculated from the information in the table.

**Discussion and Practice**

1. Calculate \( E(X) \) for the pet example.

2. Suppose that it costs around $20 per week to feed a pet. We now want to study the probability distribution for a new random variable \( Y \), the cost of pet feeding per week.

3. a. Write the probability distribution for \( Y \) in table form, based on the information provided above on number of pets per student.

4. Use the results of Problem 2a to calculate \( E(Y) \), the expected weekly pet feeding cost per student.

5. Another way to find the expected value of \( Y \) is to observe that \( Y = 20X \).

6. Use the formula for \( E(Y) \) and \( E(X) \) to show that \( E(Y) = 20E(X) \).

7. Calculate \( E(Y) \) by using the result in Problem 3a.

8. Compare this result with your answer to Problem 2b.

9. Show that, in general, if \( Y = aX + b \) then \( E(Y) = aE(X) + b \).

10. Begin by writing the formula for \( E(Y) \) as a summation, assuming \( Y \) can take on values \( y_1, y_2, \ldots, y_k \) with respective probabilities \( p_1, p_2, \ldots, p_k \).

11. Substitute \( y_i = ax_i + b \) inside the summation.

12. Use the Distributive Property to write the terms inside the summation as a sum of two terms.

13. Use the properties of sums to write the summation as a term involving \( E(X) \).
6. Cost = 500 + 100Y;
   \[ E(Cost) = 500 + 100E(Y) = 500 + 100(2.58) = 758 \]

7. Time = 10 + 30C;
   \[ E(Time) = 10 + 30E(C) = 10 + 30(0.91) = 37.30 \text{ minutes} \]

8. Possible answer: A game is fair if the expected loss for the player is zero. This implies that the expected gain for the person running the game is also zero.

9. You should be willing to pay $0.50. Then, your gain is \(-0.50\) with probability \(\frac{199}{200}\) and 99.50 with probability \(\frac{1}{200}\) for an expected gain of
   \[ E(G) = -0.50(\frac{199}{200}) + 99.50(\frac{1}{200}) = 0 \]

Practice and Applications

6. Refer to Problem 13 in Lesson 2. Suppose the cost to the Nielsen Company for connecting a family to their system is a flat rate of $500 per household plus $100 for every family member in the household. How much should Nielsen expect to pay per family for connection charges?

7. Refer to Problems 8 and 9 in Lesson 2. The Gallup organization wants to sample children under the age of 18 and ask them about their attitudes toward school. It cannot sample children directly but it can sample families. It takes about 10 minutes to question the family about the status of their children and about 30 additional minutes for each interview conducted. How much time, on the average, should Gallup allow for each family sampled?

Fair Games

In the town raffle, a drawing is to take place for a radio worth about $100. Two hundred tickets will be sold for $1 each. The tickets are mixed in a drum and one ticket is randomly selected for the winning prize. If you buy one ticket, let's analyze what happens to \(G\), the amount you gain.

There are two possible outcomes you win or you lose. If you lose you have lost $1, which can be called a gain of \(-1\). If you win, however, you gain $100 minus the $1 you paid to play, for a net gain of $99. So the probability distribution for \(G\) is as shown in the table.

<table>
<thead>
<tr>
<th>(G)</th>
<th>(P(G))</th>
</tr>
</thead>
<tbody>
<tr>
<td>-1</td>
<td>(\frac{199}{200})</td>
</tr>
<tr>
<td>99</td>
<td>(\frac{1}{200})</td>
</tr>
</tbody>
</table>

By the rules of expected value, your expected gain is

\[ E(G) = -1(\frac{199}{200}) + 99(\frac{1}{200}) = -\frac{1}{2}. \]

You can expect to lose a half dollar for every play of such a game. Would you call this a fair game?

8. Write a reasonable definition of a fair game.

9. What would you be willing to pay for a drawing like the one above to make the drawing fair in the sense of expected gain?

16 Lesson 3
For a fair game, the expected gain is zero for both the player and the operator of the game. This implies that the operator would make no money in the long run and, hence, could not stay in business. Commercial games, therefore, cannot be completely fair to the player.

J.J., a.

\[ P(W) = \begin{array}{c|c}
199 & 200 \\
\hline
0 & 100 \\
\end{array} \]

b. \( E(W) = 0(199/200) + 100(1/200) = 0.50 \)

c. \( G = W - 1.00 \)

d. \( f(G) = E(W) - 1.00 = 0.50 - 1.00 = -0.50 \), or a loss of $0.50.

The answer agrees with that of the earlier discussion. \( W \) has a slightly simpler probability distribution, with an expected value that is easy to compute mentally. Writing \( G \) as a function of \( W \) may be the easier method.

**Practice and Applications**

12. a. 

\[ W \quad P(W) \\
0 \quad \frac{N-1}{N} \\
A \quad \frac{1}{N} \\
\]

b. \( E(W) = 0(\frac{N-1}{N}) + A(\frac{1}{N}) = \frac{A}{N} \)

c. \( E(G) = E(W) - C = \frac{A}{N} - C; \) if \( E(G) = 0 \), then \( C = \frac{A}{N} \).

d. The expected amount won is the total amount available divided equally among all the tickets sold.

For a fair game, a player should be willing to pay the average amount to be won per ticket.

13. a. The amount the company pays out can be thought of as the amount "won" by the client in this "game." Denote this by \( W \).

\[ W \quad P(W) \\
0 \quad 0.94 \\
10,000 \quad 0.05 \\
20,000 \quad 0.01 \\
\]

\( E(W) = 10,000(0.05) + 20,000(0.01) = 700 \)

b. \( E(Gain \ for \ the \ company) = r - E(W) = r - 700 \)

\( E(Gain \ for \ the \ company) = r - 700 = 0 \) implies \( r = 700 \). The company has a certain cost of doing business and, therefore, cannot stay in business unless its expected gain from the policies it sells is positive.
SUMMARY

Many practical applications of probability involve finding expected values of functions of random variables. For linear functions of the form $Y = aX + b$ for constants $a$ and $b$,

$$E(Y) = aE(X) + b.$$
LESSON 4

The Standard Deviation as an Expected Value

Materials: none
Technology: graphing calculators (optional)
Pacing: 1–2 class periods

Overview
The standard deviation is introduced as a measure of variability for data displayed in a relative-frequency table, with immediate extension to the use of standard deviation as a measure of variability for a probability distribution. The variance is introduced as the average of the squared deviations from the mean, and the standard deviation is then the square root of the variance. Students are asked to develop a general expression for the variance using summation notation.

Teaching Notes
As in previous lessons, this lesson is set up so that the exercises can be completed by small groups. Have students work through the spread-sheet approach to the calculation of variance and standard deviation so that they can see the steps and understand what standard deviation measures.

Whether the divisor of the sample standard deviation should be \( n \) or \( (n - 1) \) is always a confusing issue. When calculating the standard deviation from relative-frequency data, the size of the data set \( n \) may be unknown. The calculation, then, as outlined here is equivalent to dividing by \( n \), should \( n \) be known. That is the correct way to calculate the standard deviation when it is used as a measure of variability in a probability distribution.

The idea that an interval of one standard deviation to either side of the mean will contain a somewhat predictable amount of the probability distribution merits some discussion. For mound-shaped symmetric distributions, this will be about 70% of the distribution. For skewed distributions, it will be more than 70%.

Technology
A graphing calculator will be a great help with these calculations. The suggested spread-sheet calculations can be done in the lists of such a calculator. Also, the standard deviation can be calculated directly by placing the values of the variable in one list and the relative frequencies, or probabilities, as decimals in another. The TI-83 does the calculation through the STAT/CALC/1-Var Stats command. If your calculator has two forms of the standard deviation built in, make sure students look at the one that is equivalent to dividing by \( n \).

Follow-Up
Have students calculate and interpret the standard deviation for other frequency or relative frequency distributions of data, or for other probability distributions.

Have students find and report on an article from research literature that makes use of standard deviation in drawing its conclusions.
The Standard Deviation as an Expected Value

Do most households have about the same number of cars, or is there a great deal of variation from household to household?

Is the number of persons per family more variable than the number of children per family?

How can you measure variation in a probability distribution?

In data analysis, once we have a measure of center, it is important to develop a measure of variation, or spread, of the data to either side of the center. One useful measure of variability is the standard deviation, a value you may have encountered in lessons on data analysis. We now develop that same measure of spread for probability distributions.

A deviation is the distance between an observed data point and the mean of the distribution of data. The average of the squared deviations has a special name, variance. The square root of the variance is called the standard deviation. The standard deviation has important practical uses in probability and statistics, some of which we will see in future lessons of this unit.

INVESTIGATE
Recall the data in Lesson 2 regarding the number of pets students have. The survey of 20 students revealed that 4 have no pets, 10 have one pet, and 6 have 2 pets. A tabular array for

Understand how to compute and interpret the standard deviation of data and probability distributions.
Solution Key

Discussion and Practice

1. a. The data points here are 0, 1, and 2. Thus, the deviations from the mean of 1.1 are -1.1, -0.1, and 0.9, respectively; answers will vary.
   b. The average of the deviations is \( \frac{4(-1.1) + 10(-0.1) + 6(0.9)}{20} = 0 \).
   c. The squared deviations are 1.21, 0.01, and 0.81, respectively.
   d. The average of the "squared deviations from the mean" is \( \frac{4(1.21) + 10(0.01) + 6(0.81)}{20} = \frac{9.8}{20} = 0.49 \).
   z. The standard deviation is the square root of the average of the squared deviations, or square root of the variance, 0.7 in this case; answers will vary.

2. The standard deviation is the square root of the average of the squared deviations, or square root of the variance, 0.7 in this case; answers will vary.

Discussion and Practice

The mean number of pets per student is 1.1. Do you remember how we determined this? The standard deviation is a special function of the variable "number of pets" and can be calculated by making use of what we learned in Lesson 2 and 3.

a. Use the following steps to find the variance of the number of pets per student:
   a. Add a column of "deviations from the mean" to the table. What would you say is a "typical" deviation?
   b. Find the average of the deviations from the mean.
   c. Add a column of "squared deviations from the mean" to the table.
   d. Find the average of the squared deviations from the mean, called the variance.
   z. The standard variation is the square root of the variance.
   Find the standard deviation of the number of pets per student. Is this number close to what you chose as a typical deviation in Problem 1a?
   a. Draw a bar graph of the data on the number of pets per student given above.
   a. Mark the mean of this distribution on the graph.
   b. Mark off a distance of one standard deviation above, that is, to the right of, the mean.
   c. Mark off a distance of one standard deviation below, that is, to the left of, the mean.
   d. What fraction of the 20 data values are inside the interval from one standard deviation below the mean to one standard deviation above the mean?

4. Sketch another bar graph, still using data values of 0, 1, and 2 but choosing frequencies which would have greater.
4. Possible answer:

For this distribution, the mean is 1.0 and the standard deviation is 0.89, larger than the standard deviation of the distribution in Question 3. This bar graph has greater spread than the one in Question 3 because of the higher frequencies at 0 and 2, with a corresponding lower frequency at 1.

5. a. 1.11
b. The mean is 0.91 children per family.

c. \[0.91 - 1.11 = -0.20\]
\[0.91 + 1.11 = 2.02\]
d. The data values 0, 1, and 2 are inside the interval marked off by the points of Problem 5c. These values comprise 90% of the possible values for number of children per family. Note that the lower boundary drops below zero and is outside the actual range of the data.

6. Possible answer: The standard deviation is a form of "average deviation from the mean." If we were to summarize the sizes of the deviations from the mean in a single number, this "average" is often a good summary value. An average can be thought of as a typical value in a data set.

7. a. \(E(C) = 0.91\) and \(SD(C) = 1.11\)
b. The numerical values are the same.
c. In Problem 5, the mean and the standard deviation describe the center and spread of the actual number of children per family in the U.S. In Problem 7a, the expected value and the standard deviation describe the anticipated center and spread of values for the number of children per family in a sample that is not yet selected.
8. Using the symbol \( \mu \) to denote \( E(X) \),
\[
\text{Variance}(X) = \sum_{i=1}^{n} (x_i - \mu)^2 p_i
\]
and \( \text{SD}(X) = \sqrt{\sum_{i=1}^{n} (x_i - \mu)^2 p_i} \).

NOTE: This is a big jump for students, as it moves from data to symbols. They may need some direction from the you at this point.

Discussion and Practice

9. a. \( E(\text{Number of cars}) = 2.17 \); 
\( \text{SD}(\text{Number of cars}) = 0.94 \)
b. \( 2.17 - 0.94 = 1.23 ; 2.17 + 0.94 = 3.11 \)
This interval covers the data values 2 and 3. About 65.2% of all households in the U.S. have a number of cars in this interval.
c. Possible answer: The average number of cars per household in the U.S. is 2.17. That implies that a sample of 100 typical households would have around 217 cars. The data on number of cars per household is concentrated from 1 to 3, which implies that most households have at least one car, but relatively few households have more than three cars.

Practice and Applications

9. Data on automobiles per family in the U.S. are given below.

<table>
<thead>
<tr>
<th>Number of Cars</th>
<th>Percent of Households</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>1.6</td>
</tr>
<tr>
<td>1</td>
<td>22.8</td>
</tr>
<tr>
<td>2</td>
<td>40.7</td>
</tr>
<tr>
<td>3</td>
<td>21.5</td>
</tr>
<tr>
<td>4</td>
<td>10.6</td>
</tr>
</tbody>
</table>

a. Calculate the expected value and standard deviation of the number of cars per household that would be expected in a random sample of households from the U.S.
b. What percent of the households have a number of cars within one standard deviation of the mean?
c. Suppose you are allowed to use only the mean and standard deviation to describe these data in a newspaper article to be read by people who are not familiar with these terms. Write such a description.

d. The distribution of the number of persons per household in the U.S. is given in the following table.

<table>
<thead>
<tr>
<th>Number of Persons per Household</th>
<th>Percent of Households</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>42</td>
</tr>
<tr>
<td>2</td>
<td>32</td>
</tr>
<tr>
<td>3</td>
<td>17</td>
</tr>
<tr>
<td>4</td>
<td>16</td>
</tr>
<tr>
<td>5</td>
<td>7</td>
</tr>
<tr>
<td>6</td>
<td>2</td>
</tr>
<tr>
<td>7</td>
<td>1</td>
</tr>
</tbody>
</table>
10. a. 

Possible answer: The distribution of the number of persons per household ranges from 1 to 7, with a concentration of values between 1 and 4; it is more spread out than the distribution of number of children per family. The latter has a concentration of values between 0 and 2. The standard deviation of the number of persons per household will be greater.

b. \( SD(\text{Number of persons per household}) = 1.39; \ SD(\text{Number of children per family}) = 1.11\)

11. Possible answer: If this additional information were available, the distributions would be more spread out, and the standard deviations would be greater.

12. Possible answer: For somewhat symmetric distributions, the interval mean \( \pm 1SD \) includes the values concentrated in the middle of the distribution, and usually includes 50% to 70% of the possible data values. (See the distributions on number of pets and number of cars.) For highly skewed distributions, the interval mean \( \pm 1SD \) includes the values concentrated at the end of the distribution with high frequencies, and usually includes more than 70% of the possible values, sometimes 90% or more. (See the distributions on number of children and number of persons.)

13. a. The underlying data, ages, is continuous rather than discrete; but it is difficult to construct a good histogram with the class summaries given here. One meaningful plot is simply a line graph across the midpoints of the age intervals. The “65 and over” class was arbitrarily cut off at 84 to make the last interval comparable to many of the others.

Possible answer:
shoes start out low and end up with high frequencies at the older ages. Jogging shoes peak in the middle age groups. Walking shoes must have the greatest mean age, whereas gym shoes must have the least. It is very difficult to tell how the standard deviations compare by looking at the distributions. Since ages for jogging shoes have a peak in the middle, their standard deviation may be the least. Ages for gym shoes seem to have the flattest distribution, and that may give them the greatest standard deviation. Another possible plot is the cumulative percent plot shown below. Here, the percents are added as we go up the age scale. For example, 50.0% of the gym-shoe buyers are age 17 or younger, and 83.1% are 44 or younger. This plot shows clearly that gym shoes have the youngest buyers and walking shoes the oldest.

b. From the table or the cumulative-percent plot, about half of the gym-shoe buyers are under the age of 17; about half of the jogging-shoe buyers are under the age of 33—something a little less than 34; and about half of the walking-shoe buyers are under the age of 52—not quite half way between 44 and 64.

c and d. Using midpoints of the intervals as the value for the age of each group, the means and the standard deviations are approximately those in the following table.

<table>
<thead>
<tr>
<th></th>
<th>Mean</th>
<th>Standard deviation</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gym shoes</td>
<td>24.86</td>
<td>19.89</td>
</tr>
<tr>
<td>Jogging shoes</td>
<td>34.79</td>
<td>17.23</td>
</tr>
<tr>
<td>Walking shoes</td>
<td>51.23</td>
<td>18.55</td>
</tr>
</tbody>
</table>

Students may get quite different answers, depending upon how they choose the midpoint of the age groups and how they cut off the oldest age group. The mean is higher than the median for gym shoes, which has the age distribution with severe skewness toward the older age groups. The mean and median are close for the other two types of shoes, but remember that the calculations here are rough approximations to the actual values.

e. Of three types of shoes—gym, jogging, and walking—that make up a large part of the sport-shoe market, gym shoes have the
youngest buyers and walking shoes the oldest. The average age of a gym-shoe buyer is approximately 25, but half of these buyers are 17 or younger. The typical purchaser of jogging shoes is about 35 years old, and these buyers are concentrated in an age range from 17 to 52. For walking shoes, the typical buyer is about 51 years old, with these buyers concentrated in an age range from 32 to 70.

14. a. \( SD(Y) = 14 \)

\[ Y = 20X; \quad SD(Y) = 20\cdot SD(X) \]
\[ = 20(0.70) = 14 \]

15. a. This demonstration involves symbol manipulation, which is a big jump from working with data. Students may need some guidance.

\[ V(Y) = \sum_{i=1}^{n} (y_i - \mu_y)^2 p_i \]
\[ = \sum_{i=1}^{n} [(ax_i + b) - (a\mu_x + b)]^2 p_i \]
\[ = \sum_{i=1}^{n} [a(x_i - \mu_x)]^2 p_i \]
\[ = a^2 \sum_{i=1}^{n} [(x_i - \mu_x)]^2 p_i = a^2 \cdot V(X), \]

where \( \mu_x \) denotes \( E(X) \) and \( \mu_y \) denotes \( E(Y) \).

From this result, we see that \( SD(Y) = \sqrt{V(Y)} = |a| \cdot SD(X) \).

b. With the number of persons in a randomly selected household denoted by \( Y \) and the cost of interviewing by \( C \), \( C = 30Y \),
\[ E(C) = 30E(Y) = 30(2.6) = \$78, \]
and
\[ SD(C) = 30SD(Y) = 30(1.4) = \$42. \]

Since $100 is only about one half a standard deviation above the expected cost, it could be exceeded fairly often.

**SUMMARY**

For distributions of data and probability distributions of random variables the center is often measured by the mean or expected value and the spread by the standard deviation. The standard deviation measures a "typical" deviation between a possible data point and the mean. Most of the data points usually lie within one standard deviation of the mean.

For probability distributions, the standard deviation can be written as an expected value of a function of the underlying random variable. This measure will be used extensively in future lessons of this module.
Solution Key

1. Possible answer: The distribution of number of persons per household has its greatest frequency at 2 persons per household, with 90% of the values concentrated from 1 through 4. The distribution is skewed toward larger families, with very few households having more than 7 persons. The distribution of number of children per family has its greatest frequency at 0 and 90% of the values are concentrated from 0 through 2. The distribution is skewed toward more children, with very few families having more than 4 children.

2. Possible answer: Suppose a random sample of households is to be selected from all U.S. households by a polling organization. Define the random variable T as the total number of persons per household among those households sampled. T will vary from household to household in the sample, but we anticipate that the distribution of values for T should look much like the table in Problem 1 after the sample is selected and measured.

3. a. \(1 - 0.25 = 0.75\)
   b. \(1 - 0.25 - 0.32 = 0.43\)
   c. \(1 - 0.25 - 0.32 = 0.43\)
   d. \(0.25 + 0.32 + 0.17 = 0.74\)
   e. \(0.32 + 0.17 + 0.16 = 0.65\)
   f. \(0.32 + 0.17 + 0.16 + 0.07 = 0.72 - 1 - 0.28\)

4. a. \(P(T \geq 2) = 0.75\)
   b. \(P(T > 2) = 0.43\)
   c. \(P(T \geq 3) = 0.43\)
   d. \(P(T \leq 3) = 0.74\)
   e. \(P(2 \leq T \leq 4) = 0.65\)
   f. \(P(T > 1 \text{ and } T < 6) = P(1 < T < 6) = P(2 \leq T \leq 5) = 0.72\)

5. According to the U.S. Bureau of the Census, the number of cars available to American households is given by the following percents.

<table>
<thead>
<tr>
<th>Number of Persons per Household</th>
<th>Percent of Households</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>25</td>
</tr>
<tr>
<td>2</td>
<td>32</td>
</tr>
<tr>
<td>3</td>
<td>17</td>
</tr>
<tr>
<td>4</td>
<td>16</td>
</tr>
<tr>
<td>5</td>
<td>7</td>
</tr>
<tr>
<td>6</td>
<td>2</td>
</tr>
<tr>
<td>7</td>
<td>1</td>
</tr>
</tbody>
</table>

Note: Households of more than 7 persons are very rare.
S. a. $E(C) = 2.17$

d. $SD(C) = 0.94$

e. $1000(2.17) = 2170$

d. $2.17 = 46.1$

e. $250(2.17) \approx 542.50$

f. $250(0.94) = 235.00$

g. $E(Cost \text{ for 1000 households}) = 1000(542.50) = 542,500$;

$SD(Cost \text{ for 1000 households}) = 1000(235) = 235,000$

No; the actual cost for maintenance by 1000 families could easily deviate from the expected cost by a standard deviation or more.

A random sample of households is to be selected to participate in a study entitled "How much do you spend on auto repairs?"

- What is the expected number of cars per randomly selected household?

- What is the standard deviation of the probability distribution for the number of cars per randomly sampled household?

- If the poll selects 1000 households, how many cars are expected to be represented in the poll?

- If the polling organization expects 1000 cars to be represented, how many households should be sampled?

- Car maintenance costs average $250 per year, not counting gas and oil. How much would a randomly selected household expect to pay annually for car maintenance?

- What is the standard deviation of the amount a randomly sampled household is to pay for car maintenance?

- Use the information in Problem 5e to determine the total amount a random sample of 1000 families would expect to pay for car maintenance in a year. Do you think this expected value will be a good approximation to the real total for the 1000 families? (HINT: Find the standard deviation of the total amount 1000 families might have to pay for car maintenance and use that value in your answer.)

- The following table shows the age distribution of residents of the United States for the years 1990 and 1996, according to the U.S. Bureau of the Census. The population figures are given in thousands. The columns labeled "proportion" show the proportions of the residents in each of the age categories.
The graph shows the proportions as a function of the midpoint of each age group, with 102.5 used as the midpoint of the oldest group.

The key difference between the two distributions is the shift that occurs from the teens through the 40s. The 1990 population has a large group in the teens, 20s, and 30s that by 1996 has moved over to the 20s, 30s, and 40s. Even in the short time of six years, significant shifts in the age distribution of the population can be seen.

The median ages are approximately 32 and 34 for 1990 and 1996, respectively. As should be the case, 1996 has the greater median.

The expected age for 1996 is approximately 35.96. The expected age for 1990 is approximately 35.26. Both of the means are greater than the respective medians, which is not surprising because of the skewness of the age distributions toward the older ages.

The mean ages for the two years are very close, since the extremes of the two distributions are very nearly identical. The medians differ a little more because of the shifting near the centers of the distributions.

The standard deviation for the 1996 data is approximately 22.4. For the 1990 data, it is approximately 22.2, again almost identical. This is a reasonably good approximation of "typical" deviations from the mean. Most of the ages are, in fact, between 36 - 22 = 14 and 36 + 22 = 58.
7. Possible answer: From the percents given in the data, we can arrive at the following percents for the four classes of income data.

<table>
<thead>
<tr>
<th>Education</th>
<th>Percent</th>
<th>Typical income</th>
</tr>
</thead>
<tbody>
<tr>
<td>Less than high school</td>
<td>10</td>
<td>14,000</td>
</tr>
<tr>
<td>High-school degree</td>
<td>33</td>
<td>20,000</td>
</tr>
<tr>
<td>Some college</td>
<td>28</td>
<td>23,000</td>
</tr>
<tr>
<td>College degree</td>
<td>29</td>
<td>43,000</td>
</tr>
</tbody>
</table>

From this table, \( E(\text{Income}) = \$26,910. \) This expected value is based on very broad categories of workers and rough approximations to typical incomes for each category. More detailed data could give quite a different expected value. For example, professionals with advanced degrees have much greater typical incomes than \$43,000 per year. The expected value could increase if these incomes were a direct part of the calculation.

7. According to Census data, about 90% of the U.S. workforce have at least a high-school education, about 37% have at least some college education, and about 29% have at least a bachelor's degree from a college or university. Suppose a typical worker without a high-school education earns about \$14,000 per year, a typical high-school graduate makes about \$20,000 per year, a typical worker with some college experience but not a bachelor's degree makes about \$23,000 per year, and a typical worker with at least a bachelor's degree makes about \$43,000 per year. Find the expected yearly income for a person randomly selected from the U.S. workforce. Explain why this expected value may be slightly different from the true mean income of the U.S. workforce.
Sampling Distributions of Means and Proportions
LESSON 5

The Distribution of a Sample Mean

Materials: 100 chips or small equal-sized pieces of paper for each group of students. Let the students put the numbers on the chips so that they can clearly understand every step of the simulation.

Technology: graphing calculators or computers (optional)

Pacing: 1 class period or more

Overview

A fundamental principle of statistics is that in random sampling the potential values of a sample mean have approximately a mound-shaped symmetric distribution centering at the population mean and with standard deviation equal to the population standard deviation divided by the square root of the sample size. This result, known as the "Central Limit Theorem," is best discovered by students through simulation. The sampling distribution of the mean can be called "normal," but details on the normal distribution are deferred until the next lesson.

Teaching Notes

Allow enough time for small groups of students to work through the simulation by hand, so that they can see the patterns in the distributions of the sample means unfold. You might have students do the simulation and collect the data in one session and then report the results at the beginning of the next; or the students could conduct the simulation outside of class. Organize the class discussion so that the sample results from the groups can be quickly collected and assembled for display. The dotplots (number-line plots) or stemplots of the collected means can be made on an overhead transparency or with a computer or graphing calculator with display capabilities.

It is important for students to see that the shape of the simulated sampling distribution gets more symmetric and mound-shaped as the sample size increases. Also, the mean of the sampling distribution does not change, but the variation decreases as the sample size increases. Point out that the theoretical formula for the standard deviation of the sampling distribution appears to work quite well. This result allows us to anticipate the distribution of possible values of a sample mean in random sampling quite precisely and is one of the most amazing results in probability and statistics.

Technology

The simulation here should be done by hand. A graphing calculator or computer may be used to help display the data and to calculate the summary statistics for the collected means.

Follow-Up

Another simulation of the sampling distribution for the sample mean can be done, using another of the distributions from earlier lessons, such as the number of cars per household, or a distribution of the students' choice. The second time through, students could generate the simulated sample means using a computer or graphing calculator.
LESSON 5: THE DISTRIBUTION OF A SAMPLE MEAN

STUDENT PAGE 35

LESSON 5

The Distribution of a Sample Mean

If you were to sample 100 families, what is the total number of children you would expect to see?

What is the distribution of potential values of the mean number of children per family in the sample of 100 families?

How will the distribution of potential values of the mean change with the sample size?

What is the average summer daytime temperature for your town? What is the average age of a student in your class? What is the average number of points scored by your basketball team during the season? What is the average time it takes you to get to school in the morning? Averages used as summary or typical numerical values are all around us. When working with data, we have seen that the arithmetic average is called the mean. When working with a probability distribution, a possible model for data yet to come, the average is called the expected value.

Much of the data we see in designed studies, such as sample surveys, comes about through random samples from specific populations. Since these data are typically reported in summary form as means, it is important that we understand the behavior of sample means that arise from random sampling.

INVESTIGATE

The A. C. Nielsen Company samples households to collect data on TV-viewing habits. For some shows, the company is particularly interested in the number of children under the age of 18.
SOLUTION KEY

DISCUSSION AND PRACTICE

1. We expect to see only 910 children in a sample of 1000, so the chance of seeing more than 1000 does not appear to be too great. Answers will vary, but students generally don't think things will deviate far from the mean or expected value.

2. Now, we expect to see 1092 children in a sample of 1200, so the chance of seeing more than 1000 appears to be quite good. In fact, it seems like a virtual certainty.

An approximation to the expected number of children per family, as calculated in earlier lessons, is approximately 0.91.

From information gathered so far, Nielsen can tell something about the expected number of children per family in a random sample of n families. What other information does the company need?

DISCUSSION AND PRACTICE

Nielsen wants to choose a sample size that produces a reasonable number of children with high probability.

1. If \( n = 1000 \), how many children would Nielsen expect to see in the sample of families? Suppose the company has a goal of seeing at least 1000 children in its survey. Do you think the probability of seeing more than 1000 children is high for a sample of 1000 families?

2. Suppose the sample size is increased to \( n = 1200 \). Will that increase the probability that Nielsen will achieve the goal of at least 1000 children in the sample? Will that probability change a great deal or very little?

In order to get specific answers to questions like those just posed, more information on the probability distribution of sample means from random samples must be developed; that is
3.-4. An example of the results from a simulation like the one described here is provided in Figure 1 of Problem 6 in the student text. The sampling distributions for the means are mound-shaped and nearly symmetric. They center at about 0.91, the expected value for the original distribution of children per family, and have variation that decreases as the sample size increases.

the goal of this lesson. This new information will be discovered through simulation.

The Sampling Distribution of a Mean

If one family is randomly selected from the U.S. population, what is the probability that the family will contain no children? One child? Four children? Our first job is to design a simulation for random sampling that will preserve these probabilities, yet allow us to look at typical sample data on children per family. This investigation can be completed most efficiently if you work in small groups and then combine data for the class.

5. Designing and conducting the simulation

a. Each group must have 100 plastic chips or small pieces of paper of equal size. Number the chips with integers 0 through 4, in the same percents given in the table for number of children per family, and place the chips into a box. You have now constructed a physical model of the probability distribution of the number of children per family. If you reach into the box of chips and randomly select one chip, what is the probability that it will show a 2? Is this the probability you want for modeling Nielsen's sampling process?

b. Select a simulated sample of 10 households. Randomly draw a series of 10 chips from the box, but be careful to replace each sampled chip before the next one is drawn. Why is it important to do the sampling with replacement?

c. What percent of the sample outcomes were zeros? What percent were ones? Is this approximately what you expected?

d. Calculate the mean number of children per household found in your random sample of 10 households.

e. Repeat the process for three more samples of 10 households each. Record and save the sample data and the values of the means. Your group should now have four samples.

f. Collect the sample means produced by all the groups in the class. Plot the collection of sample means on a dotplot or stemplot so that the shape of the distribution can be seen. Comment on this shape. Does it differ from the
LESSON 5: THE DISTRIBUTION OF A SAMPLE MEAN

5. The mean of each simulated sampling distribution should be close to 0.91. The standard deviation of each sampling distribution should be close to 1.114, the standard deviation of the original distribution, divided by the square root of the sample size. This result is demonstrated in Problem 6 in the student text.

You might ask students to investigate the relationship between the standard deviation of the sampling distribution and the sample size graphically. By plotting observed SD against sample size as in the first plot below, we can see that we have a type of inverse relationship. With a little investigation, students can discover that a straight line is produced by plotting observed SD against the inverse of the square root of the sample size as in the second plot below. The straight line here has a slope a little greater than 1, about the size of the population standard deviation.

STUDENT PAGE 38

shape of the population distribution of children per family? How does it differ?

4. Changing the sample size

a. Repeat the simulation of Problem 3 for samples of 20 households each. This can be accomplished by making two pairs of the size-10 samples already selected, yielding two size-20 samples. Collect the sample means produced by all groups in the class, plot the means on a dotplot or stem-and-leaf plot, and comment on the shape.

b. Repeat the simulation of Problem 3 for samples of 40 households each. This can be accomplished by combining the two size-20 samples already available in each group. Collect the sample means produced by all groups in the class, plot the means on a dotplot or stem-and-leaf plot, and comment on the shape.

c. Repeat the simulation of Problem 3 for samples of 80 households each. This can be accomplished by combining the 40 data points from your group with 40 from another group. Make sure each group's data is used only once. Collect the sample means produced by the class, plot the means on a dotplot or stem-and-leaf plot, and comment on the shape.

d. The distribution of possible values of the sample mean is called a sampling distribution. Study the plots of sampling distributions for size-10, size-20, size-40, and size-80 samples, and comment on the:
   i. shapes of the sampling distributions of sample means.
   ii. centers of the sampling distributions of sample means.
   iii. variation in the sampling distributions of sample means.

5. Computing summary statistics

a. Calculate the means and the standard deviations for each of the simulated sampling distributions produced above. That is, use the original sets of sample means generated by the class in the simulations to calculate the mean and the standard deviation of the sample means within each sample size.
b. How do the calculated means of the simulated sampling distributions compare to the expected number of children per family (the mean of the population) calculated from the population distribution to be \( \mu = 0.91 \)? (The symbol \( \mu \) used for the population mean is the Greek \( \mu \), or \( \mu \).) Make a general statement about how the means of the sampling distributions relate to the mean of the population from which the samples were selected.

c. How do the calculated standard deviations of the simulated sampling distributions compare to the standard deviation of the population, calculated to be \( \sigma = 1.114 \)? (The symbol \( \sigma \) used for the population standard deviation is the Greek \( \sigma \), or \( \sigma \).) Do you see a pattern developing in how the standard deviations of the sampling distributions relate to the sample sizes?

d. The precise relationship between the standard deviation of a sampling distribution for means and the sample size is difficult to see intuitively, so we'll provide some help. We denote the population standard deviation by \( \sigma \) and label the standard deviation of a sampling distribution of sample means by \( SD(\text{mean}) \). You have noticed that \( SD(\text{mean}) \) decreases as the sample size \( n \) increases. Mathematical theory of statistics says that the precise relationship among these quantities is given by

\[
SD(\text{mean}) = \frac{\sigma}{\sqrt{n}}
\]

For \( \sigma = 1.114 \), as it is for the population of number of children per household, calculate the theoretical \( SD(\text{mean}) \) for sample sizes of 10, 20, 40, and 80. Compare these theoretical values to the observed standard deviations of the sampling distributions calculated in Problem 5a.

**Patterns**

To show that the patterns generated in sampling distributions simulated by your class are not accidental, and to show a slightly larger simulation, we have produced a computerized version of the simulation outlined above. In this case, we have calculated the means from 100 random samples in each simulated sampling distribution. The results are shown in the dotplots that follow.
Notice three things:

- The sampling distributions all center around the population mean of 0.91.
- The standard deviations of the sampling distributions get smaller as the sample size increases.
Practice and Applications

6. a. For a sample of 10 families to contain at least 30 children, the sample mean would have to exceed 3. From the first graph in the Summary following Problem 5, we see that this never happens.

b. For a sample of 20 families to contain at least 30 children, the sample mean would have to exceed 1.5. From the second graph in the Summary, we see that this could happen but has only a small chance.

c. For a sample of 40 families to contain at least 30 children, the sample mean would have to exceed 0.75. From the third graph in the Summary, we see that this happens with probability greater than 0.5.

d. For a sample of 80 families to contain at least 30 children, the sample mean would have to exceed 0.375. From the fourth graph in the Summary, we see that this happens virtually all the time.

7. a. In a random sample of 1000 families, the potential values of the sample mean would show a mound-shaped, symmetric distribution centering at 0.91, the expected number of children per family, and with a standard deviation of 0.035. The latter implies that more than half of the time the sample means would lie between 0.875 and 0.945. So, the distribution of potential values of the sample mean is concentrated very tightly around the expected value.

b. With the distribution of sample means showing an expected value of 0.91 and a standard deviation of 0.035, it follows that a mean of 1 child per sampled family is 2.57 standard deviations above the mean of the distribution. A value greater than a point this far above the center of the distribution has a very small chance of occurring.
e. With the distribution of sample means showing an expected value of 0.91 and a standard deviation of 0.035, it follows that a mean of $\frac{1000}{1200} = 0.83$ child per sampled family is more than 2 standard deviations below the mean of the distribution. A value greater than a point this far below the center of the distribution has a very great chance of occurring; in fact, such a value will occur nearly every time.

b. We need to find a mean weight for the random sample of $n$ people who enter the elevator on any given run and choose $n$ to make sure this mean does not exceed $\frac{2000}{n}$ very often.

Suppose $n = 14$. Then $\frac{2000}{14} = 142.8$, and the standard deviation of the distribution of sample means is $\frac{10}{\sqrt{n}} = 2.67$. Since this value for the sample mean is below the expected value of the distribution of sample means (given as 150) it will be exceeded with probability greater than 0.5. This $n$ is too large for safe running of the elevator.

Suppose $n = 13$. Then $\frac{2000}{13} = 153.8$, and the standard deviation of the distribution of sample means is $\frac{10}{\sqrt{n}} = 2.77$. Since this value for the sample mean is about 1.37 standard deviations above the expected value of the distribution of sample means (given as 150) it will be exceeded with probability much less than 0.5. The weight limit will be exceeded on occasion, but this may allow for safe running of the elevator.

Suppose $n = 12$. Then $\frac{2000}{12} = 166.7$, and the standard deviation of the distribution of sample means is $\frac{10}{\sqrt{n}} = 2.89$. Since this value for the sample mean is about 5.6 standard deviations above the expected value of the distribution of sample means (given as 150) it will be exceeded almost never. This would be a safe way to run the elevator, but it may be too conservative, as the elevator may then be forced to make more trips than necessary.

9. Possible answer: Those who choose to call a radio show are not a random sample of the population at large. Thus, there is no way to tell how far this mean may actually be from the true population mean. The methods of this lesson will not work with this type of non-random data.
10. The mean of the sampling distribution of sample means should always be equal to the mean of the population from which the samples are selected, regardless of the sample size.

11. The standard deviation of the sampling distribution of sample means is equal to the population standard deviation divided by the square root of the sample size. This implies that the spread of the sampling distribution decreases as the sample size increases.

means have three important properties:

- The sampling distributions all center around the population mean.
- The standard deviations of the sampling distributions get smaller as the sample size increases, and this can be predicted by the rule:

\[
\sigma_{\text{mean}} = \frac{\sigma}{\sqrt{n}}
\]

- The sampling distributions tend to be symmetric and mound-shaped.

The first two properties were investigated in this lesson; the third is the subject of the next lesson. This result is commonly known as the "Central Limit Theorem."
LESSON 6

The Normal Distribution

Materials: random-number generator
Technology: graphing calculators or computers (optional)
Pacing: 1 class period

Overview

The normal distribution is the most common probability model used in statistics because it is the approximate sampling distribution of sample means and sample proportions. This lesson deals with its basic properties, including the fact there is a relationship between intervals on both sides of the mean and the relative frequency with which values will fall into those intervals. The mean is a very useful summary statistic, and the exercises are designed to show how knowledge of the mean's sampling distribution can lead to intelligent decisions in a variety of contexts.

Teaching Notes

Another simulation is introduced here, this time with random digits, to ensure that students grasp the idea that the normality of the sampling distribution of sample means is a general result. Students should see that the relative frequencies associated with the normal distribution allow us to state that the sample mean will be within two standard deviations of the population mean about 95% of the time. This is a very useful result for making practical decisions, as in the elevator problem.

Technology

Since the simulation of this lesson is based on random digits, it can be done with a graphing calculator or computer.

Follow-Up

The lesson deals with only the basic properties of the normal distribution and looks at probabilities in intervals of one or two standard deviations to either side of the mean. Since most graphing calculators have the normal distribution built in, this can be generalized to consider problems of any fractional distance to either side of the mean.
LESSON 6: THE NORMAL DISTRIBUTION

STUDENT PAGE 44

LESSON 6

The Normal Distribution

What is the chance that your school's mean weekly earnings from recycling aluminum cans during the fall will exceed $130?

Do the probability distributions of potential values of a sample mean always have nearly the same shape?

How can you make use of a common model for distributions of means?

OBJECTIVES

Understand the basic properties of the normal distribution.
See the usefulness of the normal distribution as a model for sampling distributions.

Means are widely used statistical summaries, and decisions based on means can be more enlightened if decision-makers understand the behavior of sample means from random samples. Suppose records show that the weekly amounts your school earns on recycling aluminum cans has a mean of $120 and a standard deviation of $8. During a sixteen-week period in the fall, what is the chance that the mean weekly earnings will exceed $130? $124? $100? In Lesson 5, you used simulation to answer questions similar to these. But it is cumbersome and time-consuming to conduct a simulation every time you want to answer a question about a potential value of a mean. It would be helpful to have a model for the behavior of sample means which would give quick approximate answers to the many questions that arise about sample means. Such a model is the normal distribution.

For the A. C. Nielsen Company, it may be important to know even more details than provided in Lesson 5 about the possible values of mean number of children per sampled family. One such question might be, "What is the chance of having fewer than 1000 children in a sample of 1200 families?" Relating the
sampling distributions discovered in Lesson 5 to a theoretical model for such distributions—the normal distribution—can help provide answers to such questions.

INVESTIGATE

Consider, once again, the plots constructed in Lesson 5 (on page 40) that show the probabilistic behavior of sample means for various sample sizes. These curves look fairly symmetric and mound-shaped. Such mound-shaped, symmetric distributions are seen very often in the practice of plotting data. In fact, they are seen so often that such a curve is called "normal," and a theoretical model for this curve has been studied extensively.

Normal curves have two key measurements that determine their location and shape. One, the location of the center of the curve on the real-number line, is the mean, usually denoted by \( \mu \). The other, the measure of spread, or width, of the curve, is the standard deviation, usually denoted by \( \sigma \). Pictures of two normal curves with different standard deviations are shown below. Note that standard deviation measures variability. A careful look at these pictures shows the standard deviation to be half of the curve about \( \frac{\sqrt{2}}{2} \) of the way up the line of symmetry, the vertical line through the mean.

The area under the curve over an interval on the horizontal axis represents the percent of the data that fall into that interval. These intervals can be located in terms of the mean and...
Solution Key

Discussion and Practice

1. a. 50%
   
   b. \( \frac{100 - 68}{2} = 16\% \)
   
   c. 34 + 47.5 = 81.5%
   
   d. 100 - 95 = 5%

2. Possible answer: No; there is a high percent of families with no children, and then the percents decrease as the number of children increases. The distribution is strongly skewed toward more children.

standard deviation of the distribution of data. For the normal curve, it is common practice to describe the distribution in terms of intervals that are symmetric about the mean. For the normal distribution,

the interval \( \mu \pm 1\sigma \) contains about 68% of the data,

the interval \( \mu \pm 2\sigma \) contains about 95% of the data,

and the interval \( \mu \pm 3\sigma \) contains about 99.7% of the data.

These intervals and their respective areas are shown in the graph that follows. The scale on the horizontal axis is in terms of standard-deviation units. A point at 1 is one standard deviation above the mean. A point at -2 is two standard deviations below the mean.

Discussion and Practice

1. For a distribution of data that can be represented by a normal curve,
   
   a. what percent of the data is below the mean?
   
   b. what percent of the data is more than one standard deviation above the mean?
   
   c. what percent of the data is between one standard deviation below the mean and two standard deviations above the mean?
   
   d. what percent of the data is more than two standard deviations away from the mean?
   
   2. Does the population distribution of number of children per family (See Lesson 3 or the Assessment for Lessons 1–4) look “normal”? Explain your reasoning.
3. Possible answer: The distributions of number of cars per household in Lesson 1 and ages of purchasers of jogging shoes in Lesson 4 are somewhat mound-shaped and symmetric, similar to a normal distribution. Students may have seen distributions of measurements on weights or heights of men or women, which tend to be approximately normal in their distribution. Standardized exam scores, like those of the SAT, tend to be normally distributed. Many other examples are possible.

4. The sampling distribution for sample means, in random samples from a fixed population, have a distribution which is approximately normal in shape, centering at the true population mean and with standard deviation equal to the population standard deviation divided by the square root of the sample size. Recall that this result is known as the Central Limit Theorem.

5. The interval \( \mu \pm \frac{\sigma}{\sqrt{n}} \) will contain about 68% of the possible values of the sample mean.
   The interval \( \mu \pm 2 \frac{\sigma}{\sqrt{n}} \) will contain about 95% of the possible values of the sample mean.
   The interval \( \mu \pm 3 \frac{\sigma}{\sqrt{n}} \) will contain about 99.7% of the possible values of the sample mean.

6. Shown on the following page are the stem-and-leaf or stemplots of the simulated sample means from Lesson 5. It is easier to count the exact number of values in prescribed intervals from this view of the data. The observed mean and the standard deviation are used in the calculation of the intervals here.

3. Recall other distributions of data you have seen in these lessons or other places. Describe at least two other data sets that look normal to you. Explain your reasoning.

**The Normal Distribution as a Model**

The task at hand is to relate the normal-looking sampling distributions for the sample mean, found in Lesson 5, to the normal distribution. The normal distribution, we discovered above, depends upon two constants, a mean and a standard deviation. Each of the sampling distributions seems to have approximately the same mean, and it is close to the population mean, or expected value, of 0.91 in the case of number of children per family. Therefore, we can take \( \mu = 0.91 \) in our theoretical normal model.

What about the standard deviation for the normal model? The standard deviation within the sampling distributions decrease as \( n \) increases, so each sample size generates a slightly different sampling distribution. These standard deviations are related to the population standard deviation by the equation

\[
\text{Standard Deviation of Sample Mean} = \frac{\sigma}{\sqrt{n}}
\]

where \( \sigma \) denotes the theoretical standard deviation of the underlying population. This relationship was explored in Lesson 5, with \( \sigma = 1.114 \) for the distribution of number of children per family.

4. Summarize the above information in a concise statement about the approximate sampling distributions of sample means for random samples of size \( n \) from a fixed population, by answering the questions below.
   a. What is the shape of the sampling distribution?
   b. Where is the center of the sampling distribution?
   c. What is the standard deviation of the sampling distribution?
   d. Rewrite the basic relative-frequency rules for the normal distribution, preceding Problem 3 above, in terms of sampling distributions for sample means.
   e. Look carefully at the distributions on page 40 of Lesson 5. What fraction of the observed sample means lie within two standard deviations of the mean of their distribution in each of the four cases? Do these fractions agree with what the normal distribution would predict?

\[
\begin{align*}
n & = 10: 0.914 \pm 2(0.3426) \text{ or } (0.229, 1.599) \text{ contains 96% of the observed data values.} \\
n & = 20: 0.932 \pm 2(0.2397) \text{ or } (0.453, 1.411) \text{ contains 95% of the observed data values.} \\
n & = 40: 0.897 \pm 2(0.1891) \text{ or } (0.519, 1.275) \text{ contains 97% of the observed data values.} \\
n & = 80: 0.916 \pm 2(0.1188) \text{ or } (0.678, 1.154) \text{ contains 94% of the observed data values.}
\end{align*}
\]

All of these percents agree closely with the 95% that the normal distribution would predict for these intervals.
LESSON 6: THE NORMAL DISTRIBUTION

Stemplot of $n = 10$  Leaf Unit = 0.010

| 1  | 1  | 0   |
| 1  | 2  |      |
| 4  | 3  | 00  |
| 6  | 4  | 00  |
| 16 | 5  | 0000000000 |
| 27 | 6  | 0000000000 |
| 33 | 7  | 000000 |
| 45 | 8  | 000000000000 |
| (10)| 9  | 000000000000 |
| 45 | 10 | 000000000000 |
| 36 | 11 | 000000000000 |
| 28 | 12 | 000000000000 |
| 18 | 13 | 000000000000 |
| 9  | 14 | 000000 |
| 4  | 15 | 000000 |
| 3  | 16 | 000000 |
| 2  | 17 | 000000 |
| 2  | 18 | 000000 |

Stemplot of $n = 20$  Leaf Unit = 0.010

| 1  | 3  | 5   |
| 2  | 4  | 0   |
| 7  | 5  | 00055 |
| 13 | 6  | 00055 |
| 27 | 7  | 00000000055555 |
| 38 | 8  | 0000005555 |
| (17)| 9  | 0000000000055555 |
| 45 | 10 | 000000000000055555 |
| 27 | 11 | 00000000000000000000 |
| 13 | 12 | 00055 |
| 8  | 13 | 005 |
| 5  | 14 | 005 |
| 2  | 15 | 00 |

Stemplot of $n = 40$  Leaf Unit = 0.010

| 1  | 5  | 2   |
| 7  | 5  | 557777 |
| 12 | 6  | 22222 |
| 14 | 6  | 77 |
| 21 | 7  | 0022222 |
| 29 | 7  | 55557777 |
| 37 | 8  | 00022222 |
| (15)| 8  | 55555555577777777 |
| 48 | 9  | 00022222 |
| 40 | 9  | 5577 |
| 35 | 10 | 00000022222 |
| 24 | 10 | 555577 |
| 18 | 11 | 000222222 |
| 8  | 11 | 55 |
| 5  | 12 | 55 |
| 5  | 12 | 55 |
| 3  | 13 | 22 |
| 1  | 13 | 5   |
7. One possible simulation of sample means for samples of size 10 from random digits is shown below. Both the dotplot and stemplot are provided, along with the summary statistics.

**DotPlot 'Means'**

<table>
<thead>
<tr>
<th>Mean</th>
<th>2.40</th>
<th>3.20</th>
<th>4.00</th>
<th>4.80</th>
<th>5.60</th>
<th>6.40</th>
</tr>
</thead>
</table>

**Stemplot of Means**

Leaf Unit = 0.10

<table>
<thead>
<tr>
<th>Mean</th>
<th>4.00</th>
<th>4.20</th>
<th>4.40</th>
<th>4.60</th>
<th>4.80</th>
<th>5.00</th>
<th>5.20</th>
<th>5.40</th>
<th>5.60</th>
<th>5.80</th>
<th>6.00</th>
<th>6.20</th>
<th>6.40</th>
</tr>
</thead>
</table>

<table>
<thead>
<tr>
<th>n</th>
<th>Mean</th>
<th>Median</th>
<th>Standard Deviation</th>
</tr>
</thead>
<tbody>
<tr>
<td>50</td>
<td>4.364</td>
<td>4.500</td>
<td>0.952</td>
</tr>
</tbody>
</table>

The distribution is mound-shaped and somewhat symmetric, but we will not see a perfect normal curve for a sample size of 10 and only 50 means in the simulation.

For random digits, the theoretical mean, or mean of the population, is 4.5 and the standard deviation is 2.872. These can be worked out by the methods of Lessons 2 and 4. The observed mean of the simulated sampling distribution is 4.364, quite close to what the theory predicts. The observed standard deviation of the simulated sampling distribution is 0.952, again quite close to \( \frac{2.872}{\sqrt{10}} = 0.908 \). The normal distribution seems to be a good model here.

8. The probability that the distance between the sample mean and the population mean is less than two standard deviations of the sampling distribution for the mean is approximately 0.95 because of the approximate normality of this sampling distribution.
LESSON 6: THE NORMAL DISTRIBUTION

Practice and Applications

9. a. $0.91 \pm 2 \frac{1.114}{\sqrt{25}}$; $0.91 \pm 0.45$ or $(0.46, 1.36)$
   b. $0.91 \pm 2 \frac{1.114}{\sqrt{100}}$; $0.91 \pm 0.22$ or $(0.69, 1.13)$
   c. $0.91 \pm 2 \frac{1.114}{\sqrt{1000}}$; $0.91 \pm 0.07$ or $(0.84, 0.98)$
   d. $0.91 \pm 2 \frac{1.114}{\sqrt{4000}}$; $0.91 \pm 0.04$ or $(0.87, 0.95)$

10. The sample total $= n \cdot \text{(sample mean)}$ and an interval for sample totals can be found by simply multiplying both ends of the appropriate interval for means by the common sample size.
   a. $(11.5, 34)$, where $34 = 25 \cdot 1.36$ and $11.5 = 25 \cdot 0.46$
   b. $(69, 113)$, where $113 = 100 \cdot 1.13$ and $69 = 100 \cdot 0.69$
   c. $(840, 980)$, where $980 = 1000 \cdot 0.98$ and $840 = 1000 \cdot 0.84$
   d. $(3480, 3800)$, where $3800 = 4000 \cdot 0.95$ and $3480 = 4000 \cdot 0.87$

11. Because this is a difficult problem, we have provided the following steps which lead to its solution.

   The point that cuts off the lower 0.025, or 2.5%, on the normal distribution is 2 standard deviations below the mean.

   If the total number of children falls below 1000, then the mean number of children per family falls below $\frac{1000}{n}$.

   The mean of the sampling distribution for sample means is 0.91.

   Therefore, we must choose the sample size $n$ so that the point $\frac{1000}{n}$ lies 2 standard deviations below 0.91, or

   $$ \frac{1000}{n} - 0.91 = 2 \frac{0.22}{\sqrt{n}} \Rightarrow n = 1183 \text{ or } 1184.$$

   12. For the sampling distribution of mean weekly earnings over a 16-week period, the standard deviation is $\frac{8}{4} = 2$.

   The $\$130$ figure is 5 standard deviations above the mean, and has virtually no chance of being exceeded.

   The $\$124$ figure is 2 standard deviations above the mean and will be exceeded with probability 0.025.

   The $\$100$ figure is 10 standard deviations below the mean and will be exceeded with probability 1.
This analysis is based on the fact that the earnings for the 16 weeks of the term will behave like a random sample from the distribution of possible earnings. This may not be the case if the fall term in question is, for instance, unusually cold so that fewer cold drinks are consumed.

13. \( n = 14 \): The value \( \frac{2000}{14} = 142.9 \) is 2.68 standard deviations below the mean of 150. Thus, it will be exceeded between 97.50% and 99.85% of the time.

\( n = 13 \): The value \( \frac{2000}{13} = 153.8 \) is 1.37 standard deviations above the mean. Thus, it will be exceeded between 2.5% and 16% of the time, with the true percent being a little closer to the 16%.

\( n = 12 \): The value \( \frac{2000}{12} = 166.7 \) is 5.79 standard deviations above the mean, and should never be exceeded.

14. The normal distribution sometimes works well as a model of the distribution of the population itself. If the population has a normal distribution, then the sampling distribution for a sample mean will be normally distributed for any sample size, even a sample size of 1. The three histograms that follow show basketball scores from all NCAA college playoff games between 1939 and 1995. Plot A is based on the score of the winner, plot B on the score of the loser, and plot C on the total points scored in the game.

What number would you use for maximum occupancy? Explain your reasoning.

What is the approximate probability that the weight limit will be exceeded if the number of people on the elevator is 14? 13? 12?

number too small, but dangerous to make it too large.

What is the approximate probability that the weight limit will be exceeded if the number of people on the elevator is 14? 13? 12?

14. The normal distribution sometimes works well as a model of the distribution of the population itself. If the population has a normal distribution, then the sampling distribution for a sample mean will be normally distributed for any sample size, even a sample size of 1. The three histograms that follow show basketball scores from all NCAA college playoff games between 1939 and 1995. Plot A is based on the score of the winner, plot B on the score of the loser, and plot C on the total points scored in the game.
14. a. Yes; an appropriate normal distribution would serve as an adequate model for all three distributions. The distribution of total scores is most nearly normal. The distribution of winner scores has a slightly extended tail on the side of the greater values, which is reasonable since there is a tendency for scores of the winners to be large. The distribution of loser scores has a slight bulge on the lower side of the mean; that is, it is not quite symmetric. This, too, is reasonable since the scores of the losers tend to be small, but not extremely so.

b. Winners' scores: mean, approximately 77; standard deviation, approximately 14
Losers' scores: mean, approximately 66; standard deviation, approximately 13
Total scores: mean, approximately 144; standard deviation, approximately 26

c. The interval $77 \pm 2(14)$ or $(49, 105)$ should include about 95% of the winners' scores. The bars on the histogram below 50 and above 105 indicate that this is approximately correct.

d. A total score of 180 is only 1.4 standard deviations above the mean. Thus, it could be exceeded fairly often, between 2.5% and 16% of the time. From the data that make up the histogram, the score of 180 was exceeded 130 times out of 1706 scores, or 7.6% of the time.
SUMMARY

Sampling distributions for sample means, the distributions of potential values of a sample mean, are used widely in making decisions with data. The normal distribution provides a good model for these sampling distributions as long as the samples are randomly selected from a fixed population. The normal distribution is characterized by its symmetric mound shape, mean, and standard deviation. From knowledge of the mean and standard deviation, relative frequencies within intervals to either side of the mean can be found.
LESSON 7

The Distribution of a Sample Proportion

Materials: physical simulation device such as spinners or numbered chips for each group of students
Technology: graphing calculators or computers (optional)
Pacing: 1–2 class periods

Overview

Sample proportions in random sampling behave in a manner that exactly parallels the behavior of sample means, and this lesson has a structure that parallels that of Lesson 5. The sampling distribution of sample proportions is approximately normal with center at the true population proportion and with standard deviation that decreases as the sample size increases according to a known formula. This amazing and useful result allows us to make practical decisions on problems that involve sample proportions. It might be a good strategy to have students conduct the simulation and collect the data in one session and report on the results in another.

Teaching Notes

Again, the approximate normality of the sampling distribution for a proportion is best discovered by simulation. It is good if at least part of the simulation is done by hand so that students can see what is happening. A calculator or computer can be used to record results and graph the data, as in the figure that accompanies the lesson. The simulation requires an event with probability 0.6. This can be accomplished with physical devices such as spinners and numbered chips. Consider some of these options before going to technology. Make sure that students understand what is meant by a proportion in the sense of working with data.

Technology

As mentioned above, graphing calculators and computers can be used to summarize the data from the simulation, but it would be good if some of the simulation could be done with physical devices. It is quite easy, however, to do this simulation with random digits generated by a calculator or computer.

Follow-Up

The lesson outlines a generic simulation. Have students find an article in the media reporting on a sample survey with results summarized in a proportion. These surveys generally report a margin of error. Ask the students to design a simulation to see if the reported margin of error appears to be correct.
The Distribution of a Sample Proportion

How many high-school graduates would you expect to see in a random sample of 1000 adults?

Is there a good chance that a sample of 1000 adults could produce over 900 high-school graduates?

Is the distribution of potential values of a sample proportion similar to the distribution of potential values for a sample mean?

What percent of the students in your school like the food in the cafeteria? What proportion of your income do you spend on food and entertainment? What fraction of the residents of your town own their homes? Data are often summarized by reporting a percent or a proportion for one or more categories involved. In order to make intelligent decisions based on data reported this way, we must understand the behavior of proportions that arise from random samples.

Investigate

The Sampling Distribution of a Proportion

First, let's be clear about the basic terminology. Suppose 20 students are asked to report their favorite food and 12 say “pizza.” The proportion reporting pizza is \( \frac{12}{20} = 0.60 \) and the percent reporting pizza is 60%. Most of the problems in this lesson will deal with proportions, which will be numbers between 0 and 1.
Solution Key

Discussion and Practice

1. a. Most students will see that 40(0.6) = 24 is a reasonable answer.
   b. Students are not expected to calculate an exact answer here. Some will reason that 30 does not seem far from the expected value of 24, so the probability might be fairly high at 0.3 or so.
   c. For many students, 10 will seem like a long way from the expected value of 24, so the probability should be small.

2. This simulation can be done quite easily with the random-number generator in a graphing calculator or computer. Statistical software packages often have routines built in to do this kind of random sampling. If you do not have access to technology of this type, random digits can be physically drawn out of a box.

   The results given in Problem 4 demonstrate one possible set of outcomes for a simulation of this type.

   - This simulation can be done quite easily with the random-number generator in a graphing calculator or computer. Statistical software packages often have routines built in to do this kind of random sampling. If you do not have access to technology of this type, random digits can be physically drawn out of a box.

   The results given in Problem 4 demonstrate one possible set of outcomes for a simulation of this type.

Discussion and Practice

Thinking About Survey Results

1. a. It is estimated that about 60% of automobile drivers use seat belts. Suppose your class is to conduct a survey of 49 randomly selected drivers. Think about the following questions. Try to arrive at reasonable answers based upon your current knowledge and your intuition. Do not spend a lot of time with calculations.
   b. How many drivers would you expect to be using seat belts?
   c. What is the chance that more than 30 drivers will be using seat belts?
   d. Would it be quite unusual to find fewer than 10 of the drivers wearing seat belts?

   Now, we will develop the tools we need to answer these questions more precisely.

   An intuitive “feel” for how sample percents behave in random sampling can be developed through simulation. Let Y denote the number of successes in a random sample of size n from a population in which the probability of success on any one sample selection is given by p. This section will concentrate on properties of the sample proportion Y/n. Y/n can be changed into a percent by multiplying by 100, but we will work most directly with the decimal form. This investigation can be completed most efficiently if you work in pairs.

   Designing and Conducting the Simulation

   a. Assume that p = 0.6 for this study.
   b. Find a device that will generate an outcome that has probability equal to 0.6 of occurring. A random-number table, a calculator that generates random numbers, or slips of paper may be used.
   c. Let the sample size be n = 10. Generate ten outcomes by the device agreed upon in Problem 2a and count the successes among the ten outcomes. Divide the number of successes by 10 to obtain the proportion of successes. That is the sample proportion that will be recorded.
   d. Repeat Problem 2b three more times so that your group has four different samples of size 10 each and four observed values of the sample proportion.
d. Combine your values of the sample proportion with those from the rest of the class. Plot the sample proportions on a dotplot or a stem-and-leaf plot.

e. Based on the results of the simulation, find approximate values for each probability.

i. \( P(0.5 \leq \hat{p} \leq 0.7) \)

ii. \( P(0.4 \leq \hat{p} \leq 0.8) \)

iii. \( P(\hat{p} > 0.8) \)

e. Calculate the mean and the standard deviation of the simulated distribution of sample proportions. Keep these for future reference.

g. From your dotplot or stem-and-leaf plot, describe the distribution of the possible values of \( \hat{p} \) for samples of size 10, when the true value of \( p \) is 0.6. NOTE: This distribution is called a simulated sampling distribution of the proportion \( \hat{p} \).

Changing the Sample Size

a. Repeat the simulation of Problem 2 for samples of size \( n = 20 \). Keep \( p \) fixed at 0.6. To save time, two of the size-10 samples may be combined to obtain a size-20 sample. Complete all parts of Problem 2.

b. Repeat the simulation of Problem 2 for samples of size \( n = 40 \). Keep \( p \) fixed at 0.6. To save time, two of the size-20 samples may be combined to obtain a size-40 sample. Complete all parts of Problem 2.

c. Compare the shapes of the simulated sampling distributions for samples of size 10, 20, and 40.

d. Compare the means of the simulated sampling distributions for samples of size 10, 20, and 40. How close are these means to the true value of \( p \)?

e. Compare the standard deviations of the simulated sampling distributions for samples of size 10, 20, and 40. A theoretical result in probability states that the standard deviation of sample proportions should be related to the true \( p \) and the sample size by the following rule:

\[
\text{SD(proportion)} = \sqrt{\frac{p(1-p)}{n}}
\]
Does this rule appear to hold for the sampling distributions generated in this investigation?

Summary of Sampling Distributions for Sample Proportions
4. Your plots may look something like those shown below. These were generated by computer for the same cases described above, with the addition of a sampling distribution for samples of size 80.

Sampling Distributions of Sample Proportions
Each dot represents 2 points.
4. **a.** All four of the sampling distributions are somewhat mound-shaped and symmetric; at least they do not have a long tail in either direction. All four center at about the same place on the real-number line. The variation in the distributions decreases as the sample size increases.

   **b.** All four of the distributions center close to 0.6, which is the population proportion of successes.

   **c.** Using the rule $SD(\text{proportion}) = \sqrt{\frac{p(1-p)}{n}}$ with $p = 0.6$, the theoretical standard deviations are as follows:
   
   - $n = 10$: $SD = 0.1549$
   - $n = 20$: $SD = 0.1095$
   - $n = 40$: $SD = 0.0775$
   - $n = 80$: $SD = 0.0548$

   These are quite close to the observed standard deviations of the sampling distributions shown in the lesson, which are 0.1552, 0.0962, 0.0838, and 0.0540, respectively.
5. a. \( n = 10: 0.60 \pm 2(0.1549) = 0.91 \) or 0.29 contains 99/100 of the data values.
   \( n = 20: 0.60 \pm 2(0.1095) = 0.82 \) or 0.38 contains 98/100 of the data values.
   \( n = 40: 0.60 \pm 2(0.0774) = 0.75 \) or 0.44 contains 96/100 of the data values.
   \( n = 80: 0.60 \pm 2(0.0547) = 0.71 \) or 0.49 contains 94/100 of the data values.

The actual data values inside these intervals are a little difficult to discern from the graph, but a close approximation is all that is needed here.

b. Yes; the four fractions are relatively close to each other. The estimate of the probability in question should be the same for each of the four sample sizes.

c. Yes; all four of the estimates of the probability of a sample proportion being within two standard deviations of \( p \) are about 0.95, as the theory states.

Practice and Applications

6. a. Possible answer: Many newspapers and news magazines such as *Time*, *Newsweek*, and *U.S. News and World Report* almost always contain at least one poll. The results for categorical questions are always reported in terms of percents.

b. Possible answer: Percents adjust for the sample size, so it makes sense to compare results for two different polls or for two different groups within a poll by looking at percents. For example, your school could be compared with a much larger or much smaller school on the question of student participation in athletics by looking at the percents of students who compete in each school. It would not be fair to compare the actual numbers of students participating in athletics.

c. Possible answer: Often the details on how the sampling was done are not given. The exact questions asked of the participants may not be given. Some of the sample sizes and percents you would like to have for analysis may not be given in the report.

7. The "margin of error" or the "sampling error" is the quantity \( 2SD(proportion) = 2\sqrt{\frac{p(1-p)}{n}} \).
   This tells you that the sample proportion is quite likely to lie within this distance of the true population proportion.
b. Possible answer: If \( p = 0.5 \) rather than 0.6, the centers of the distributions would shift to 0.5 but the shapes would remain about the same. If \( p = 0.8 \) rather than 0.6, the centers of the distributions would shift to 0.8. The shapes would become more skewed toward the smaller values, especially for small sample sizes, since the proportions are constrained to be between 0 and 1.

c. \( P(\text{high-school graduate}) = 0.8 \)

d. \( 1000(0.8) = 800 \)

e. If \( n(0.2) = 400 \), then \( n = \frac{400}{0.2} = 2000 \)

d. Possible answer: No; the chance of seeing a non-graduate is small (0.2) and, therefore, a large sample would be required to have a good chance of seeing 500 non-graduates. But, because of the random nature of the sample, there is no guarantee that a sample of any size, no matter how large, would produce at least 500 non-graduates.

e. For \( n = 400 \) and \( p = 0.8 \), the standard deviation of the sampling distribution of sample proportions is 0.02. The sample proportion of \( \frac{335}{400} = 0.84 \) is 2 standard deviations above the mean of the sampling distribution. Thus, the chance of seeing more than 335 graduates in a sample of 400 is about 0.025.

f. The expected value is \( 40(0.6) = 24 \).

g. For samples of size 40 and \( p = 0.6 \), the standard deviation of the sampling distribution of sample proportions is 0.0774. The sample proportion of \( \frac{36}{40} = 0.75 \) is 1.94 (nearly 2) standard deviations above the mean of the distribution and, therefore, the chance of seeing more than 30 drivers using seat belts is a little over 0.025.

h. Speculate as to what would happen to the shapes, means, and standard deviations of the simulated sampling distributions of in Problem 4 if \( p \) were changed to 0.5. What if \( p \) were changed to 0.8?

i. According to the U.S. Bureau of the Census, about 80% of U.S. residents over the age of 25 are high-school graduates. A Gallup poll is to be conducted among those over the age of 25—we'll call these "adults" for now—for issues surrounding education. So the Gallup organization wants to be sure there will be adequate numbers of both high-school graduates and non-graduates in the resulting random sample of adults.

a. What is the probability that a randomly selected adult is a high-school graduate?

b. If 1000 adults are randomly sampled, how many high-school graduates would you expect to see?

c. If Gallup expects to have 400 non-graduates in the sample, how many adults should be randomly selected?

d. Can Gallup be assured of at least 500 non-graduates in a poll of any fixed size?

e. If Gallup randomly samples 400 adults, is it likely that the poll will result in more than 335 high-school graduates? Explain.

f. Here is a restatement of the first set of questions posed in this lesson. Answer them as specifically as you can with the information gained in this lesson.

It is estimated that about 60% of automobile drivers use seat belts. Suppose your class is to conduct a survey of 40 randomly selected drivers. Think about the following questions. Try to arrive at reasonable answers based upon your current knowledge and your intuition. Do not spend a lot of time with calculations.

a. How many drivers would you expect to be using seat belts?

b. What is the chance that more than 30 drivers will be using seat belts?

c. Would it be quite unusual to find fewer than 10 of the drivers wearing seat belts?
SUMMARY

Sample proportions or percents are used often to summarize data on categorical variables, like gender, ethnic group, attitude on public issues, and health condition. In order to make decisions based on this kind of data, we must know something about the anticipated behavior of sample proportions from random samples. What we have discovered is that sample proportions have approximately normal sampling distributions, with mean equal to the true population proportion \( p \) for the characteristic being studied and standard deviation given by the following rule:

\[
SD(\text{proportion}) = \sqrt{\frac{p(1-p)}{n}}
\]
**ASSESSMENT: LESSONS 5-7**

**Solution Key**

1. A sampling distribution for a sample mean shows the distribution of the possible values of the sample mean that could occur in a random sample from a fixed population. The sampling distribution is the distribution of potential values of the mean that could be produced by a random sample, thus giving the investigator some idea of what might happen when a random sample is actually taken. The underlying population distribution shows the distribution of the data values in a population before any sampling is done. A random sample of measurements is selected from the population distribution to produce the mean on which the sampling distribution is based.

Sampling distributions depend upon random sampling and describe potential values of a statistic. Population distributions describe the actual values a variable can take on.

2. a. \( n = 5 \): the sampling distribution of the sample mean is approximately normally distributed with mean 2.58 and standard deviation 0.62, although for samples this small the sampling distribution may retain some of the features of the population distribution.

   b. \( n = 100 \): the sampling distribution of the sample mean is approximately normally distributed with mean 2.58 and standard deviation 0.14.

   c. \( n = 1000 \): The sampling distribution of the sample mean is approximately normally distributed with mean 2.58 and standard deviation 0.04.

   The sampling distributions will look more and more normal as the sample size increases.

3. To achieve this goal, \( \frac{1.39}{\sqrt{n}} \).

   a. \( n = 10 \) or \( \sqrt{n} = 10 \).

   b. \( n = 100 \) or \( \sqrt{n} = 10 \).

   c. \( n = 1000 \) or \( \sqrt{n} = 10 \).

4. The interval \( σ = 2.58 \pm \frac{1.39}{\sqrt{1200}} \) or \( 2.58 \pm 0.04 \) yields the interval (2.54, 2.62), which means that the interval should contain the middle 68% of possible values of the sample mean.

5. If the total is to be at least 2000, then the sample mean must be at least \( \frac{2000}{n} \), where \( n \) is the sample size. The upper 97.5% of the normal distribution falls above a point which is 2 standard deviations below the mean of the sampling distribution, 2.58 in this case. Thus, we have the equation \( \frac{2000}{n} - 2.58 < -2 \frac{1.39}{\sqrt{n}} \) or \( 2.58n - 2.78 \sqrt{n} - 2000 = 0 \).
This equation can be solved by trial and error or with a graphing calculator or computer to yield $n = 806$. 

6. The distance between the sample mean and the population mean is less than 2 standard deviations of the sampling distribution with probability 0.95. Therefore, $K$ must equal 2 standard deviations, or $K = 2 \frac{\sigma}{\sqrt{n}} = 2 \frac{1.39}{\sqrt{1200}} = 2(0.04) = 0.08$.

The assumption is that the 1200 households must be selected randomly from the population of U.S. households.

7. The distribution of possible values of the sample proportion is approximately normal in shape, with center at $p$ and with standard deviation given by $\sqrt{\frac{p(1-p)}{n}}$.

8. The interval containing the middle 95% of the potential sample proportions is $p \pm 2 \sqrt{\frac{p(1-p)}{n}}$

$$0.6 \pm 2 \sqrt{\frac{0.6(0.4)}{100}}$$

$$= 0.6 \pm 2(0.049)$$

or $(0.502, 0.698)$

9. The margin of error is 2 standard deviations of the sampling distribution. Thus, the sample size $n$ must satisfy $2 \sqrt{\frac{0.6(0.4)}{n}} = 0.04$

$$\sqrt{n} = 2 \sqrt{\frac{0.6(0.4)}{0.04}}$$

$$= \frac{4(0.6)(0.4)}{(0.04)^2} = 600.$$

10. a. With $p = 0.3$ and $n = 500$, the standard deviation of the sampling distribution is 0.02. A sample proportion of 0.4 is 5 standard deviations above the mean of the sampling distribution, and so it would be very unusual to see such a result.

b. This would be a very unusual result under the usual conditions of good sampling practice. There are three possible explanations for such a result:

The sample may not have been selected randomly.

The population figure of 30% without health insurance could be wrong.

Everything was done properly and the population figure is correct, but the investigator obtained a result that would very rarely occur.
11. a. The female heights have a mean of about 65 inches and a standard deviation of about 2 inches. The mean is nearly at the center of this somewhat symmetric distribution, but is pulled a little lower by the long tail to the left. The standard deviation is half the thickness of the curve about $\frac{2}{3}$ of the way up the center line, the line of symmetry.

b. The male heights have a mean of about 70 inches and a standard deviation of a little more than 2 inches, say about 2.5 inches. The reasoning is the same as in Problem 11a.

c. A height 1 standard deviation above the mean will satisfy this condition, which gives $65 + 2 = 67$ inches.

d. A height one standard deviation below the mean will satisfy this condition, which gives $70 - 2.5 = 67.5$ inches.
Two Useful Distributions
LESSON 8

The Binomial Distribution

Materials: none
Technology: graphing calculators or computers (optional)
Pacing: 2 class periods, with exercises assigned as homework

Overview
The approximate normality of the distribution of sample proportions works well only when the sample size is reasonably large. In the case of small samples, we must use a more precise result for the probability of the number of “successes” in a random sample of size \( n \). The probabilities that result from a simple model for this situation turn out to be the terms in a binomial expansion.

Teaching Notes
This is the most mathematical of the lessons so far. Students may need to be reminded of some of the basic rules of probability, such as the multiplication rule for independent events and the addition rule for mutually exclusive events. It is important for them to see how the mathematics of probability comes together to build a very useful model.

Students can profit from group work in the development of the binomial model, but should work through some of the exercises on their own.

Technology
Some of the binomial probabilities, for small sample sizes, should be calculated by hand so that students can see each component part. After that experience, binomial probabilities can be calculated automatically on most graphing calculators or statistics software packages.

Follow-Up
Have students compare the exact binomial probabilities they get from their calculators with the normal approximations from earlier lessons. Have them try this for sample sizes of 5, 10, 50, and 100 to see that the normal approximation is good for large samples but perhaps not so good for small ones. Note that the normal approximation is stated for proportions, so that students will have to translate from proportions to counts.
The Binomial Distribution

In a random sample of four adults, what is the probability of sampling three high-school graduates?

How do small sample distributions differ from large sample distributions?

Of what use are mathematical models for calculating probabilities?

You learned in Lesson 7 that sample proportions tend to have normal sampling distributions; so the normal probability model works well for approximating probabilities associated with sample proportions. However, the normal approximations work well only in cases for which the sample size is large and the population proportion is not close to zero or 1. For small samples or for cases in which is far from 0.5, we need a more precise model.

INVESTIGATE

Sampling Distributions

It is reported that about 80% of adults over 25 in the U.S. are high-school graduates. Suppose a Gallup poll randomly selects only four adults from a small city for which this population percent is thought to hold. The first plot that follows shows a simulated sampling distribution for the sample proportion of high-school graduates in such a sample. The second plot shows the simulated distribution of the actual count of the number of high school graduates per sample of size 4.

OBJECTIVES

Understand the basic properties of the binomial distribution.

Use the binomial distribution as a model for certain types of counts.
**Solution Key**

**Discussion and Practice**

1. These distributions are highly skewed toward the smaller values. They look nothing like normal distributions, and could not be approximated well by normal distributions.

---

**Discussion and Practice**

1. Describe the shapes of the two distributions plotted. Do they look as if they could be approximated well by normal distributions?

We will not attempt to give a general rule as to when the normal distribution is an appropriate approximation to the sampling distribution for a proportion; when in doubt, set up a small simulation to see if you think the normal model will work well.

**A Random Variable for a Categorical Outcome**

In the Gallup poll mentioned above, the outcome for any one sampled person is either "high-school graduate" or "not a high-school graduate." What can we do with these categorical outcomes to change them into numerical ones? One way to make the transition from categorical to numerical is to assign a 1 if we see the characteristic we are looking for and to assign a 0 if we do not see that characteristic. We might, for example, define "high-school graduate" to be of primary interest in the Gallup poll. Thus, for the first person sampled, we would record a 1 if that person is a high-school graduate and a 0 if not. Symbolically, we could write $X_1 = 1$ if sample person 1 is a...
2. a. \( E(X_1) = 0(1 - p) + 1(p) = p \)
   
b. It is easier to work with the variance, the square of the standard deviation, at first. Let \( V(X) \) denote the variance of a random variable \( X \). Then,
   
   \[ V(X_1) = E((X_1 - p)^2) \]
   
   \[ = (0 - p)^2(1 - p) + (1 - p)^2(p) \]
   
   \[ = (1 - p)(p^2 + (1 - p)^2) \]
   
   \[ = (1 - p)p = p(1 - p) \]
   
   and \( SD(X_1) = \sqrt{p(1 - p)} \)

3. The probability distribution for \( X_2 \) is exactly like that for \( X_1 \).
   
   \[ \begin{array}{cc}
   X_2 & P(X_2) \\
   0 & 1 - p \\
   1 & p \\
   \end{array} \]

high-school graduate and \( X_1 = 0 \) if sample person 1 is not a high-school graduate.

The random variable \( X_1 \), then, completely describes the outcomes of interest for the first person sampled. The probability distribution for \( X_1 \) can be written as

\[ X_1 \begin{cases} 
1 & \text{if } X_1 = 1 \\
0 & \text{if } X_1 = 0 \\
\end{cases} \]

where \( p \) represents the probability of selecting an item with the characteristic of interest. In the high school graduate example, \( p = 0.8 \).

b. a. Find the expected value of \( X_1 \). How does it relate to the probability of sampling a high-school graduate?
   
b. Find the standard deviation of \( X_1 \). Write this as a function of \( p \).
   
c. Suppose \( X_2 \) represents the outcome for a second adult sampled from the city, coded in the same way as \( X_1 \). Give the probability distribution for \( X_2 \). (HINT: The second adult is selected independently from the first.)

The Binomial Distribution

We are ready to develop a formula for the probability of obtaining a number \( Y \) of high-school graduates in a sample of \( n \) adults from a city in which \( p \) is the proportion of adults which are, in fact, high-school graduates. We begin by letting \( n = 2 \). In this case, \( Y \) can equal 0, 1, or 2; there are no other possibilities. \( Y \) is 0 if neither of the selected adults is a high-school graduate. Symbolically,

\[ P(Y=0) = P(X_1 = 0 \text{ and } X_2 = 0) = P(X_1 = 0) \cdot P(X_2 = 0) = (1 - p)^2 \]

Multiplication is valid here since you are looking at the intersection of two independent events. For the sake of convenience in writing, let \( (1 - p) = q \). \( Y \) is 1 if exactly one of the adults is a high-school graduate. Symbolically,

\[ P(Y=1) = P(X_1 = 1 \text{ and } X_2 = 0) + P(X_1 = 0 \text{ and } X_2 = 1) \]

\[ = P(X_1 = 1) \cdot P(X_2 = 0) + P(X_1 = 0) \cdot P(X_2 = 1) \]

\[ = pq + qp = 2pq \]

THE BINOMIAL DISTRIBUTION 69
Letting $Y$ denote the total number of successes in a random sample of three adults:

\[ P(Y = 0) = P(X_1 = 0 \text{ and } X_2 = 0 \text{ and } X_3 = 0) \]
\[ = P(X_1 = 0) \cdot P(X_2 = 0) \cdot P(X_3 = 0) \]
\[ = (1 - p)^3 = q^3 \]

\[ P(Y = 1) = P(X_1 = 1 \text{ and } X_2 = 0 \text{ and } X_3 = 0) \]
\[ \text{ or } (X_1 = 0 \text{ and } X_2 = 1 \text{ and } X_3 = 0) \]
\[ = p \cdot q \cdot p + q \cdot p \cdot q + q \cdot q \cdot p \]
\[ = 3pq^2 \]

\[ P(Y = 2) = P(X_1 = 1 \text{ and } X_2 = 1 \text{ and } X_3 = 0) \]
\[ \text{ or } (X_1 = 0 \text{ and } X_2 = 1 \text{ and } X_3 = 1) \]
\[ = p \cdot p \cdot q + q \cdot p \cdot p + q \cdot q \cdot p \]
\[ = 3p^2q \]

\[ P(Y = 3) = P(X_1 = 1 \text{ and } X_2 = 1 \text{ and } X_3 = 1) \]
\[ = p^3 \]

\[ E(Y) = 0(q^3) + 1(3pq^2) + 2(3p^2q) + 3(p^3) \]
\[ = 3pq^2 + 2p^2q + 3p^3 \]
\[ = 3p(q + p) \]
\[ = 3p \text{ since } (q + p) = 1. \]

The expected value of 3.2 indicates the average number of high-school graduates that would appear in many random samples of size 4. Suppose 100 different polls each randomly selected four adults from the population. These 100 samples should average 3.2 high-school graduates per sample, for a total of 3.2(100) = 320 high-school graduates in all polls combined.

Why can we add probabilities on the second line? Finally, $Y$ is 2 if both adults are high-school graduates. Symbolically, $P(Y = 2) = P(X_1 = 1 \text{ and } X_2 = 1) = p^2$.

By combining the above results, you can represent the probability distribution for $Y$ in the case $n = 2$ with the following table.

<table>
<thead>
<tr>
<th>$Y$</th>
<th>$P(Y)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>$q^3$</td>
</tr>
<tr>
<td>1</td>
<td>$3pq$</td>
</tr>
<tr>
<td>2</td>
<td>$3p^2q$</td>
</tr>
</tbody>
</table>

4. Show that the probability distribution of $Y$ for the sample size $n = 3$ is given by the following table.

<table>
<thead>
<tr>
<th>$Y$</th>
<th>$P(Y)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>$q^3$</td>
</tr>
<tr>
<td>1</td>
<td>$3pq^2$</td>
</tr>
<tr>
<td>2</td>
<td>$3p^2q$</td>
</tr>
<tr>
<td>3</td>
<td>$p^3$</td>
</tr>
</tbody>
</table>

From the probability distributions, we can find general expressions for the expected value and standard deviation of $Y$. We will look at the specific cases for $n = 1$, $2$, and $3$ and see what generalization this suggests. For the expected value, or mean, of the probability distribution, we have:

\[ n = 1 \quad E(Y) = E(X_1) = 0q + 1p = p \]
\[ n = 2 \quad E(Y) = E(X_1) + E(X_2) = 2p(q + p) = 2p \]
\[ n = 3 \quad E(Y) = 3p \text{ when } n = 3. \]

The obvious guess for a general result is that $E(Y) = np$ for distributions of this type, and that is correct. Thus, in a random sample of four adults from the city being studied, we would expect to see about $4(0.8) = 3.2$ high-school graduates.

5. The expected number of high-school graduates is not an integer. Provide a meaningful interpretation of this number.

When developing expressions for standard deviations, it is easier to begin with the variance, the square of the standard deviation, and then take a square root at the end. We denote variance by $V$, and recall that the variance is the expected value

\[ V(Y) = \sum (Y_i - E(Y))^2 P(Y_i) \]
\[ = (0 - p)^2 q^3 + (1 - p)^2 3pq^2 + (2 - p)^2 3p^2 q + (3 - p)^2 p^3 \]
\[ = 3p^2q + 2pq^2 + pq + q \]
\[ = 3p(q + p) \]
\[ = 3p \text{ since } (q + p) = 1. \]
7. As in Problem 2, it is easier to begin with the variance.
\[
V(Y) = E(Y - 2p)^2
\]
\[
= (0 - 2p)^2(q^2) + (1 - 2p)^2(2pq)
\]
\[
+ (2 - 2p)^2(p^2)
\]
\[
= 4p^2(q^2) + (1 - 2p)^2(2pq)
\]
\[
+ 4q^2(p^2)
\]
\[
= 2pq[4pq + (1 - 2p)^2]
\]
\[
= 2pq(4pq + (1 - 4p + 4p^2))
\]
\[
= 2pq(4p - 4p^2 + 1 - 4p + 4p^2)
\]
\[
= 2pq(1)
\]
\[
= 2pq
\]

8. Using the formula 
\[
SD(Y) = \sqrt{np(1-p)}
\]
with \( p = 0.8 \), it follows that the standard deviations for the respective sample sizes are as follows.
- \( n = 4 \): 0.80
- \( n = 10 \): 1.26
- \( n = 20 \): 1.79
- \( n = 40 \): 2.53

The standard deviations of the total number of high-school graduates in the samples increase as the sample size increases. This is quite different from the behavior of standard deviations of sample proportions, in which the standard deviation decreases as the sample size increases. This is why it is easier to accurately estimate a proportion than a total, based on random samples.

9. These are simple binomial expansions. If multiplication of algebraic expressions such as these has not been covered elsewhere, this question could take a little time to develop.
10. We show the answer for the case $n = 3$. The case $n = 2$ is a simpler version. For $n = 3$, $c$ can take on the values 0, 1, 2, and 3.

$P(Y = 0) = \frac{3!}{0!3!} p^0(1 - p)^3$

$= (1 - p)^3 = q^3$

since $0! = 1$ by definition.

$P(Y = 1) = \frac{3!}{1!2!} p^1(1 - p)^2$

$= 3p(1 - p)^2 = 3pq^2$

$P(Y = 2) = \frac{3!}{2!1!} p^2(1 - p)^1$

$= 3p^2(1 - p)^1 = 3p^2q$

$P(Y = 3) = \frac{3!}{3!0!} p^3(1 - p)^0$

$= p^3$

Note that $c! = c(c-1)(c-2) \ldots (1)$ and $0! = 1$, which may require some explanation if students do not recall this fact.

### Practice and Applications

11. a. The binomial probability distribution for the case $n = 4$ and $p = 0.8$ is as follows:

<table>
<thead>
<tr>
<th>$Y$</th>
<th>$P(Y)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.0016</td>
</tr>
<tr>
<td>1</td>
<td>0.0256</td>
</tr>
<tr>
<td>2</td>
<td>0.1536</td>
</tr>
<tr>
<td>3</td>
<td>0.4096</td>
</tr>
<tr>
<td>4</td>
<td>0.4096</td>
</tr>
</tbody>
</table>

Students should be encouraged to use a calculator or computer to obtain these results.

12. a. In this case, the sample size is 5 and the probability of success on any one question is 0.5, since the student is guessing among two possible answers. The probability distribution of $Y$, the number of correct answers, is:

$c. E(Y) = np = 4(0.8) = 3.2$

d. $SD(Y) = \sqrt{4(0.8)(0.2)} = 0.80$

e. The interval mean plus and minus 1 standard deviation covers the interval $(2.4, 4.0)$. The values of $Y$ that lie within this interval are 3 and 4, and the probability that either of these values will occur is $2(0.4096) = 0.8192$. Note that this probability is much greater than the probability of being within 1 standard deviation of the mean under the normal distribution.
(12 a).

<table>
<thead>
<tr>
<th>$Y$</th>
<th>$P(Y)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.03125</td>
</tr>
<tr>
<td>1</td>
<td>0.15625</td>
</tr>
<tr>
<td>2</td>
<td>0.31250</td>
</tr>
<tr>
<td>3</td>
<td>0.31250</td>
</tr>
<tr>
<td>4</td>
<td>0.15625</td>
</tr>
<tr>
<td>5</td>
<td>0.03125</td>
</tr>
</tbody>
</table>

b. Getting at least 60% of the questions correct implies that the student must guess three or more answers correctly. $P(Y \geq 3) = 0.50$, the sum of the last three entries of the table. This answer can be seen by realizing that this distribution is symmetric.

13. For a ten-question true-false test, with a student guessing on every answer, the probability distribution for $Y$, the number of correct answers, is:

<table>
<thead>
<tr>
<th>$Y$</th>
<th>$P(Y)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.000977</td>
</tr>
<tr>
<td>1</td>
<td>0.009766</td>
</tr>
<tr>
<td>2</td>
<td>0.043945</td>
</tr>
<tr>
<td>3</td>
<td>0.117188</td>
</tr>
<tr>
<td>4</td>
<td>0.205078</td>
</tr>
<tr>
<td>5</td>
<td>0.246094</td>
</tr>
<tr>
<td>6</td>
<td>0.205078</td>
</tr>
<tr>
<td>7</td>
<td>0.117188</td>
</tr>
<tr>
<td>8</td>
<td>0.043945</td>
</tr>
<tr>
<td>9</td>
<td>0.009766</td>
</tr>
<tr>
<td>10</td>
<td>0.000977</td>
</tr>
</tbody>
</table>

A student must guess 6 or more answers correct in order to get at least 60% on this test. The probability of 6 or more correct answers is only 0.377, the sum of the last five entries on the table. The student has a better chance of obtaining a passing grade by guessing on the test with the fewer questions.

14. a. The chance of at least one donor out of ten with type-B blood is $1 - (0.9)^{10} = 1 - 0.349 = 0.651$. The complement of “at least one” is “none.” This can be solved directly with a calculator, but it can also serve as a good exercise in the use of logarithms.

b. Two ways of finding an answer to this problem have been presented in this module. The direct way is to use the binomial distribution with $n = 100$ and $p = 0.1$. A graphing calculator or computer can handle sample sizes this large. The exact sample calculation gives $P(Y \geq 10) = 0.549$.

An alternative is to use the normal approximation to the distribution of proportions. The sample proportion of donors with type-B blood in a random sample of size 100 has approximately a normal distribution with a mean of 0.1. Since the normal distribution is symmetric with respect to its mean, the chance of seeing a sample proportion larger than 0.1 is approximately 0.5.
c. As in Question 14b, there are two approaches to this problem. From direct binomial calculations, 
\( P(Y \geq 16) = 0.04 \). Using the normal approximation for sample proportions, as outlined in Question 14b, requires the standard deviation, 0.03. Thus, 0.16 is exactly 2 standard deviations above the mean of the distribution, which is 0.10. Under the normal distribution, a randomly selected value lies more than 2 standard deviations above the mean with probability 0.025. This is not quite the same as the exact binomial value, but it is not a bad approximation to use in situations in which the technology to calculate the exact value may not be readily available. In either case, the blood bank does not have a good chance of getting the 16 type-B donors out of a group of 100 donors.

This result is a simple application of the fact that 
\( E(cY) = cE(Y) \) and 
\( SD(cY) = cSD(Y) \) for any positive constant c. These ideas were developed in earlier lessons.

16. a. Since the median income has half of the incomes below it and half above it, the chance of a randomly selected income exceeding the median is 0.5. So, the basis for the probability in question is a binomial distribution with \( n = 5 \) and \( p = 0.5 \). Letting \( Y \) denote the number of households with income exceeding the median, the probability distribution is as given above, right:

<table>
<thead>
<tr>
<th>( Y )</th>
<th>( P(Y) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.03125</td>
</tr>
<tr>
<td>1</td>
<td>0.15625</td>
</tr>
<tr>
<td>2</td>
<td>0.31250</td>
</tr>
<tr>
<td>3</td>
<td>0.31250</td>
</tr>
<tr>
<td>4</td>
<td>0.15625</td>
</tr>
<tr>
<td>5</td>
<td>0.03125</td>
</tr>
</tbody>
</table>

The probability that four or more households have incomes exceeding the median, among five randomly sampled households, is 0.1875, the sum of the lowest two entries on the table. NOTE: It is good to have students get into the habit of looking at the whole distribution when answering specific probability questions of this type.

b. i. If \( X \) denotes the number of households in a random sample of 16 with incomes below the median, then \( X \) has a binomial distribution with \( n = 16 \) and \( p = 0.5 \). Thus, 
\( E(X) = np = 16(0.5) = 8 \).

ii. The standard deviation of \( X \) is given by 
\( SD(X) = \sqrt{np(1-p)} = \sqrt{16(0.5)(0.5)} = 2 \).

iii. A quick, approximate answer for the probability of seeing at least 10 of the 16 households with incomes under the median comes by observing that 10 is exactly 1 standard deviation above the mean of 8. With \( p = 0.5 \), the binomial distribution will be mound-shaped and symmetric, so that normal approximations should be good. The probability of seeing a data value more than 1 standard deviation above the mean under the normal distribution is 0.16. The exact binomial calculation gives 
\( P(X \geq 10) = 0.227 \). Notice that the normal approximation did not work as well for \( n = 16 \) as it did for \( n = 100 \).

c. This would be a very rare result, and you might suspect that the sample was not selected randomly. The sample selection method seems biased toward households with larger incomes.
### Lesson 8: The Binomial Distribution

17. a. \( P(\text{at least one alarm sounds}) = 1 - P(\text{all alarms fail}) = 1 - (0.9)^n = 0.973 \)

b. \( 1 - (0.3)^6 = 0.99927; \) no, the two probabilities are nearly the same.

c. \( 1 - (0.3)^3 = 0.99 \) has as its closest solution \( n = 4. \) This equation can be solved by trial and error or by logarithms.

18. a. If \( Y \) denotes the number of defective tapes in the sample, then \( P(Y \geq 1) = 1 - P(Y = 0) = 1 - (1 - p)^n, \) where \( p \) is the probability of observing a defective tape. This probability is to be 0.5. With \( p = 0.1, \) the equation becomes \( 1 - (0.9)^n = 0.5, \) which has \( n = 7 \) as its closest solution.

b. With \( p = 0.05, \) the equation is \( 1 - (0.95)^n = 0.5, \) which has \( n = 14 \) as its closest solution.

19. a. If the firm's gain is denoted by \( G \) and the number of wells that produce oil is denoted by \( Y, \) then \( G = 1,000,000Y - 500,000, \) since it costs \$50,000 to drill each of ten wells. Then, \( E(G) = 1,000,000E(Y) - 500,000 = 1,000,000(10)(0.1) - 500,000 = 500,000. \)

b. \( SD(G) = 1,000,000SD(Y) = 1,000,000 \sqrt{10(0.1)(0.9)} = 948,663 \)

c. Yes; there is a good chance the firm will lose money since 0 is only about one half a standard deviation below the expected gain. Yes; there is a good chance that the firm can make \$1.5 million, since this figure is only about 1 standard deviation above the expected gain.

20. If \( Y \) denotes the number of CD players that fail in the first month, \( Y \) has a binomial distribution with \( n = 10 \) and \( p = 0.08. \) The gain \( G \) for the seller is \( G = 100(10) - 200Y, \) since the seller sells all ten players for \$100 each. It follows that \( E(G) = 100(10) - 200(10)(0.08) = 1000 - 160 = 840. \)
for $100 each with a "double your money back" guarantee if the CD player fails in the first month of use. Suppose the probability of such a failure is 0.08. What is the expected net gain for the seller after all 10 CD players are sold? Ignore the original cost of the CD players to the seller.

**SUMMARY**

For a random sample of size $n$ from a population with probability of "success" $p$ on each selection, the probability that the number of "successes" in the sample $Y$ equals a specific count $x$ is given by

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

where $\binom{n}{x}$ is the binomial coefficient given by

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

The expected value of $Y$ and the standard deviation of $Y$ are given, respectively, by

$$E(Y) = np$$

and

$$SD(Y) = \sqrt{npq}$$

The random variable $Y$ is said to have a binomial distribution.
LESSON 9

The Geometric Distribution

Materials: 200 pennies for each group
Technology: graphing calculators or computers (optional)
Pacing: 1–2 class periods

Overview

Whereas the number of successes in a fixed number of independent trials leads to the binomial distribution, the sequential number on which the first success occurs leads to the geometric distribution. The development of the geometric distribution uses the same building blocks as the binomial, namely, a sequence of independent trials in which all trials have the same probability of success. The geometric is the simplest of the waiting-time distributions and has many applications in that area.

Several examples of waiting-time situations are counting the days until it rains or until the sun shines, counting the minutes until the telephone rings, counting the times you have to take an exam until you pass, and counting the number of extra points made by a place-kicker in football until he misses. Note in the latter example that success does not have to be a positive occurrence; it is simply the outcome you are looking for in the counting process.

Teaching Notes

This distribution has an infinite number of possible outcomes and, hence, derivation of its properties requires work with infinite series. Thus, it may be appropriate to review some basic ideas of infinite series, or introduce them if the students have not seen them before.

There are two simulations introduced in this lesson. The first is similar in spirit to the ones for the binomial distribution and can be done with either random digits or a physical device such as a spinner. The second requires sets of pennies and should not be programmed for the computer or calculator. Discuss with the students how these simulations differ and how they are similar. Both lead to the geometric distribution.

This lesson can be completed in one or two class periods if both simulations are done by all groups. An alternative is to assign one simulation to the groups as homework and then have students report the results in class.

Technology

Once the geometric distribution is developed, the computations are fairly easy and do not require much technology. The first simulation can be done with the graphing calculator or computer, but can also be easily accomplished by hand.

Follow-Up

There are many applications of the geometric distribution, such as the coupon-collecting problem in the assessment for Lessons 8 and 9. Have students find similar contests or games and ask them to use the geometric distribution to model how long it might take them to win.
LESSON 9: THE GEOMETRIC DISTRIBUTION

STUDENT PAGE 76

LESSON 9

The Geometric Distribution

What is the probability that your team will have to play four games before it gets its first loss?

What is the general shape of waiting-time distributions?

How can a mathematical model be developed for waiting-time distributions?

OBJECTIVES

Understand the basic properties of the geometric distribution.

Use the geometric distribution as a model for certain types of counts.

THE OPENING EXAMPLE OF LESSON 8 CONSIDERED THE PROBLEM OF HOW MANY HIGH-SCHOOL GRADUATES SHOULD APPEAR IN A RANDOM SAMPLE OF FOUR ADULTS FROM A CITY. SUPPOSE, HOWEVER, THAT THE INTERVIEWER IS INTERESTED IN FINDING ONLY ONE HIGH-SCHOOL GRADUATE FROM THIS CITY. A QUESTION OF INTEREST IN THIS CASE MIGHT BE, "HOW MANY PEOPLE MUST BE SELECTED UNTIL A HIGH-SCHOOL GRADUATE IS FOUND?" IN OTHER WORDS, "HOW LONG WILL WE HAVE TO WAIT TO FIND THE FIRST SUCCESS?" THE FIRST PERSON SELECTED MIGHT BE A HIGH-SCHOOL GRADUATE BUT, IF NOT, A SECOND PERSON WILL HAVE TO BE INTERVIEWED, AND SO ON.

INVESTIGATE

How Long Must We Wait?

Describe another situation in which waiting time until "success" is an important consideration.

Discussion and Practice

1. Recall that the proportion of high-school graduates in the population is 0.80.

76 LESSON 9.
LESSON 9: THE GEOMETRIC DISTRIBUTION

Solution Key

Discussion and Practice

1. a. No; theoretically, the sampling could go on for a long time before the first high-school graduate is found, even though there is an 80% chance of finding a high-school graduate on any one selection.

b. Yes; in light of the answer to Problem 1a, the sample size required to find the first high-school graduate could be quite large.

c. No; since the chance of finding a high-school graduate on the first selection is 0.8, the chance of finding at least one high-school graduate among the first few selections should be quite large.

2. a-e. Various devices can be used for generating this simulated data, including random numbers or colored chips (8 red and 2 white). The data displayed in this answer were produced with the random-number generator in a computer. In a total of 99 trials, success was found on the first trial 81 times, on the second trial 15 times, and on the third trial 3 times. In this simulation, it never took more than 3 trials to find the first successful outcome. The estimated probabilities for the three observed outcomes are simply the relative frequencies with which they occurred.

<table>
<thead>
<tr>
<th>Y</th>
<th>Frequency</th>
<th>P(Y)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>81</td>
<td>0.818182</td>
</tr>
<tr>
<td>2</td>
<td>15</td>
<td>0.151515</td>
</tr>
<tr>
<td>3</td>
<td>3</td>
<td>0.030303</td>
</tr>
</tbody>
</table>

The distribution of the values of Y are pictured in the graph below. The distribution is highly skewed toward the higher values, as it should be for a distribution in which a small value is quite likely to occur but a large value has only a small chance of occurring.

The mean of this distribution is a little greater than 1, being pulled in that direction by the long tail toward the larger values. Actually, the mean for this distribution is 1.21.
LESSON 9: THE GEOMETRIC DISTRIBUTION

3. **a-d.** This activity should be done with real pennies, not with a computer or calculator simulation. The computer or calculator can be used to keep track of the data.

The data below are the result of one such simulation with 200 pennies.

<table>
<thead>
<tr>
<th>Y</th>
<th>Frequency</th>
<th>P(Y)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>102</td>
<td>0.510</td>
</tr>
<tr>
<td>2</td>
<td>48</td>
<td>0.240</td>
</tr>
<tr>
<td>3</td>
<td>28</td>
<td>0.140</td>
</tr>
<tr>
<td>4</td>
<td>11</td>
<td>0.055</td>
</tr>
<tr>
<td>5</td>
<td>3</td>
<td>0.015</td>
</tr>
</tbody>
</table>

Of the 200 coins, 102 turned up heads on the first toss. Of the 98 tails, 48 turned up heads on the second toss, and so on. Again, the probabilities for the values of Y are estimated by the relative frequency of outcomes out of the 200 coins. Thus, it is estimated from the data that the probability that a "fly" will die in the first day is 0.51, quite close to the theoretical value of 0.50 and the probability of dying in the second day is 0.24. Note that there are some "flies" that lived longer than five days.

The graph of the simulated probability distribution for Y, the age of the flies in days, is shown below. This distribution is highly skewed toward more days, which is to be expected. The mean will be pulled upward from 1 quite strongly by the long tail to the right. In this case, the mean, or estimated expected value of Y, is 1.79.

Remember, however, that there were 8 "flies" still living when the simulation was stopped, and these would increase the mean even more.

**STUDENT PAGE 78**

1. Construct a plot to represent the simulated distribution of Y.
2. Describe the shape of this distribution.
3. Approximate the mean of this distribution.

**3.** Each fly of a certain species has a 0.5 chance of dying in any one-day period after its birth. Simulate the probability distribution of Y, the age of flies of this species. In this simulation, "age" can be thought of as the number of days until death, or the waiting time until death. Instead of generating the values of Y one at a time, as we did in Problem 2, consider a simulation that gives a whole set of Y values simultaneously.

**a.** Gather a set of about 200 pennies. Place the pennies in a container, shake well, and toss them onto a table. Count the heads.

**b.** If "head on a coin" represents "death of a fly" what fraction of flies died during the first day? Does this seem like a reasonable estimate of the probability that a fly died during the first day, Y = 1? Explain.

**c.** Gather the pennies that came up tails on the first toss. Place them in the container, shake well, and toss them onto a table. Count the heads. What is the approximate probability that a fly dies during the second day, Y = 2?

**d.** Continue gathering the tails, tossing them, and counting the heads until there are five or fewer tails left.

1. Record the values of Y and the approximate probability for each.
2. Show the simulated probability distribution for Y on an appropriate graph.
3. Describe the shape of this distribution.
4. Approximate the mean of this distribution.

4. **4.** The two methods of simulation are quite similar and either could be used on the first problem. The event "fly dies when two days old" is essentially the same type of event as "it takes two selections to find the first high-school gradu-

ate." The problem in using the second method (coins) on the first problem is that you would have to find something that can be selected out with probability 0.8 at each stage. Ten-sided dice will work, for example.

THE GEOMETRIC DISTRIBUTION 89
5. a. The shapes are similar in that all three are highly skewed toward the greater values. The shapes differ in that there is a greater probability at 1 and a more pronounced skewness as $p$ increases.
b. Although difficult to see on the graphs, the probabilities do not quite add to 1, since there is always a small chance that the number of trials until the first success could be greater than any value yet observed.
c. The means are approximately 3.0, 2.0, and 1.2, respectively.
6. \( a. \) \( P(Y = 2) = P(X_1 = 0) \cdot P(X_2 = 1) = q \cdot p \)  
   \( P(Y = 3) = P(X_1 = 0) \cdot P(X_2 = 0) \cdot P(X_3 = 1) = q^2 \cdot p \)  

\( b \) and \( e. \) \( P(Y = c) = P(X_1 = 0) \cdot P(X_2 = 0) \cdot \ldots \cdot P(X_{c-1} = 0) \cdot P(X_c = 1) = q^{c-1} \cdot p \)  

for any positive integer \( c. \) For \( Y \) to take on the value \( c, \) the first \( c - 1 \) trials must be failures followed by the first success on trial numbered \( c. \)
**LESSON 9: THE GEOMETRIC DISTRIBUTION**

(6) **d.** The probabilities of the successes or failures on the individual trials are multiplied together. This is allowed only if the trials are independent of one another. Independence would not be the case, for example, if names were to be selected without replacement among the students in the classroom until the first female is selected. "Without replacement" means that each selected name is kept out of the pool while the next one is drawn.

7. **a.** Following the notation used on the probability distribution of \( Y \),

\[
E(Y) = 1(p) + 2(qp) + 3(q^2p) + 4(q^3p) + \ldots
\]

Factoring out a \( p \) and writing the terms in a triangular array gives the answer.

---

**STUDENT PAGE 81**

- Using the result of part b, write \( P(Y = c) \) in terms of \( p \) and \( q \). Make the result as compact as possible.

- What assumption is necessary for the probability calculations used in part c?

With the answer to Problem 6, you can now see what the probability distribution for the random variable \( Y \) looks like in table form.

<table>
<thead>
<tr>
<th>( Y )</th>
<th>( P(Y) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>( p )</td>
</tr>
<tr>
<td>2</td>
<td>( qp )</td>
</tr>
<tr>
<td>3</td>
<td>( q^2p )</td>
</tr>
<tr>
<td>4</td>
<td>( q^3p )</td>
</tr>
<tr>
<td>\ldots</td>
<td>\ldots</td>
</tr>
</tbody>
</table>

The sum of these probabilities form an infinite series, and the sum of the probabilities across all possible values of \( Y \) form an infinite sum:

\[
\sum_{n=0}^{\infty} q^n p = \frac{p}{1-q} = \frac{p}{1-q} = (1 + q + q^2 + q^3 + \ldots) = \frac{p}{1-q} = 1.
\]

The infinite sum inside the brackets is called a geometric progression. The basic result for summing such a series is as follows:

\[
s + as + as^2 + as^3 + \ldots = \frac{s}{1-s}
\]
as long as the absolute value of \( s \) is less than 1.

You see, now, why the probability distribution developed in this lesson is called the geometric distribution.

7. The expected value of the geometric distribution can be found by summing an array of infinite series. Show that \( E(Y) \), for the geometric random variable \( Y \), can be written in a triangular array as

\[
E(Y) = \sum_{n=0}^{\infty} q^n p = \frac{p}{1-q}
\]

- Sum each row of the array inside the brackets by using the result for geometric progressions.
b. The sums for the first four rows in the triangular array, as shown in the diagram, are \( \frac{1}{p} \), \( \frac{q}{p} \), \( \frac{q^2}{p} \), \( \frac{q^3}{p} \) and with similar terms to follow in the infinite array.

c. \[ E(Y) = p \left( \frac{1}{p} + \frac{q}{p} + \frac{q^2}{p} + \frac{q^3}{p} + \ldots \right) \]
\[ = 1 + q + q^2 + q^3 + \ldots \]
\[ = \frac{1}{1 - q} = \frac{1}{p} \]

d. The expected number of trials until the first success is the reciprocal of the probability of success on any one trial. Thus, if you toss a balanced coin, you would expect to take two tosses to see the first head. If you make only 40\% of your free throws in basketball, you would expect to wait an average of 2.5 throws until your first successful attempt, if the throws are independent.

b. a-c.

<table>
<thead>
<tr>
<th>( p )</th>
<th>0.3</th>
<th>0.5</th>
<th>0.8</th>
</tr>
</thead>
<tbody>
<tr>
<td>( E(Y) )</td>
<td>3.33</td>
<td>2.00</td>
<td>1.25</td>
</tr>
<tr>
<td>( SD(Y) )</td>
<td>2.79</td>
<td>1.41</td>
<td>0.56</td>
</tr>
<tr>
<td>Interval</td>
<td>(0.54, 0.59, 0.69, 6.12)</td>
<td>(0.31, 0.41, 1.81)</td>
<td></td>
</tr>
</tbody>
</table>

The table shows the values of \( p \) for the three graphs of Figure 1 and then provides the expected values, standard deviations, and the interval of one standard deviation to either side of the expected value. Note that for small values of \( p \) the standard deviation and the interval are quite large and do not provide much guidance as to where most of the values of \( Y \) would lie. The standard deviation is a useful guide to where the values of \( Y \) are concentrated only for situations involving a large value of \( p \).

STUDENT PAGE 82

b. Find \( E(Y) \) by summing the row totals you just found, again using the result for geometric progressions.

d. The expected wait for two graduates is twice the wait for one, because of the reasoning of Problem 9c. Thus, the answer is 1.25 + 1.25 = 2.50.

9. a. \( P(Y = 3) = (0.2)^3 (0.8) = 0.032 \)
b. \( P(Y \geq 3) = 1 - P(Y = 1) - P(Y = 2) = 1 - 0.8 - 0.2(0.8) = 0.04 \)
c. The answer is the same as for Problem 9b since the process of waiting starts over once the first success is found. The trials remain independent with a constant probability of 0.8 for success.
d. \( E(Y) = \frac{1}{\overline{p}} = \frac{1}{0.8} = 1.25 \)

10. a. You are looking for non-defective engines. Therefore, \( p = 0.9 \), the probability of selecting a non-defective engine.
\[ P(Y = 3) = (0.1)(0.1)(0.9) = 0.009 \]
b. \( E(Y) = \frac{1}{0.9} = 1.11 \)
(10) c. $SD(Y) = \sqrt{\frac{0.1}{0.9}} = 0.35$

d. Letting cost be denoted by C, C = 100Y, since it costs $100 to test each engine. Thus,
$E(C) = 100E(Y) = 100(1.11)$
$= $111 and $SD(C) = 100SD(Y)$
$= 100(0.35) = $35.
The figure of $200 is about 2.5 standard deviations above the mean, and a point that far away from the mean will not be reached often.

11. a. If the lines are busy 60% of the time, the chance of getting a call through is only $p = 0.4$. Thus, $P(Y = 1) = 0.4$, $P(Y = 2) = 0.6(0.4) = 0.24$, and $P(Y = 4) = (0.6)^3(0.4) = 0.0864$.

b. $E(Y) = \frac{1}{0.4} = 2.5$

12. a. You are looking for a successful well, and $p = 0.1$. Thus,
$E(Y) = \frac{1}{0.1} = 10$.

b. Since each well costs $50,000 to drill, the cost of drilling C is given by $C = 50,000Y$. Thus,
$E(C) = 50,000E(Y)$
$= 50,000(10)$
$= $500,000 and
$SD(C) = 50,000SD(Y)$
$= 50,000(\sqrt{0.9})$
$= $474,342.
Notice that the SD is almost as large as the expected cost. There is great variability in this process.

c. The probabilities for these “at least” statements can be worked out directly, but it is useful to develop a general formula for such problems. In general,
$P(Y \geq c) = q^{c-1}p + q^{c}p + q^{c+1}p + ...$
$= p\left(q^{c-1} + q^{c} + q^{c+1} + ...ight)$
$= p\left(\frac{q^{c-1}}{1-q}ight) = q^{c-1}$.

SUMMARY

For a sequence of selections from a large population in which the probability of "success" $p$ stays the same for all selections, the number of the selection on which the first success occurs $Y$ has a geometric distribution. The mean and the standard deviation of this distribution are given by
$E(Y) = \frac{1}{p}$
and
$SD(Y) = \frac{1}{p}\sqrt{\frac{1}{1-p}}$.
ASSESSMENT

Lessons 8 and 9
Solution Key

1. a. Let \( Y \) denote the number of donors with type-A blood. For this case, \( p = 0.42 \) and 
\[ E(Y) = 5(0.42) = 2.1. \]
The probability that a random donor has type-A or type-AB blood is \( 0.42 + 0.04 = 0.46 \). The expected number of donors, out of the five, with type-A or type-AB blood is \( 5(0.46) = 2.3. \)
b. For \( n = 5 \) and \( p = 0.42 \), the distribution of the number of successes is shown below:

\[
\begin{array}{c|c}
Y & P(Y) \\
\hline
1 & 0.06564 \\
2 & 0.23765 \\
3 & 0.34418 \\
4 & 0.24923 \\
5 & 0.01307 \\
\end{array}
\]

This probability distribution is fairly symmetric, but is slightly skewed toward the greater values, since \( p \) is slightly below 0.5.
c. Let \( Y \) denote the number of type-A donors in a sample of 200 with \( p = 0.42 \). Then,
\[ E(Y) = 200(0.42) = 84 \] and \( SD(Y) = \sqrt{200(0.42)(0.58)} = 7.0. \)
d. The 100-donor goal is 2.3 standard deviations above the mean under a distribution which would look nearly normal. A value this far away from the mean has a very small chance of occurring.
e. \( P(\text{number of trials until first success is greater than } c) = (1 - p)^c. \)
For the case of type-A blood, \( p = 0.42 \), and the probability of waiting through more than four donors for the first success is \( (0.58)^4 = 0.1132. \)
For the case of type-AB blood, \( p = 0.04 \), and the probability of waiting through more than four donors for the first success is \( (0.96)^4 = 0.8493. \)
f. \( E(\text{waiting period for a type-AB donor}) = \frac{1}{0.04} = 25. \)
The blood bank could wait quite a long time to find a donor with a rare blood type.

2. a. For counting \( Y \), the number of O-donors, the probability of success is \( p = (0.46)(0.15) = 0.069 \) because of the independence of the two conditions. For \( n = 5 \), the distribution of \( Y \) is given at right:

\[
\begin{array}{c|c}
Y & P(Y) \\
\hline
1 & 0.699437 \\
2 & 0.038419 \\
3 & 0.002847 \\
4 & 0.000106 \\
5 & 0.000002 \\
\end{array}
\]
This probability distribution is extremely skewed toward the greater values because of the very small chance of success on any one randomly selected donor.
If the blood bank needs 80 A-positive donors this week, does it stand a good chance of getting them? Explain.

The blood bank is in need of an A-positive donor as soon as possible. Describe the distribution of the number of donors that must be tested to find the first A-positive donor. What is the expected number of donors to be tested in order to find one who is A-positive?

You are to take a ten-question multiple choice exam. Each question has four choices, of which only one is correct. You know none of the answers and decide to guess at an answer for each question.

Describe how to set up a simulation for the probability distribution of the number of questions answered correctly.

What is the probability that you will answer at least 60% of the questions correctly?

Suppose the teacher changes the questions so that each one has three choices, one of which is correct. Would your chance of answering at least 60% correctly by guessing go up or down from the answer on Problem 3b? Explain.

Refer to the original scenario of four choices per question. Suppose the test now has five questions rather than ten. Does your chance of getting at least 60% correct by guessing increase or decrease from your chance in Problem 3b? Explain.

Each box of a certain brand of cereal contains a coupon that can be redeemed for a poster of a famous sports figure. There are five different coupons, representing five different sports figures.

Suppose you are interested in one sports figure in particular. Set up a simulation that would produce an approximate probability distribution for the number of boxes of cereal you would have to buy in order to get one coupon for that particular poster. You stop buying cereal when you get the poster you want.

What is the probability that you would get the coupon for the particular sports figure of interest in four or fewer boxes of cereal? You stop buying cereal when you get the poster you want.

Possible answer: The probability of a correct choice is \( p = \frac{1}{4} = 0.25 \).

Random digits can be used to simulate an event with this probability. One way to do it is to disregard two of the digits, say the 9 and 0, and use 1 and 2 to indicate success among digits sampled from 1 through 8. Another randomization model is to toss two coins and let success be denoted by the occurrence of two heads. After a randomization device is selected, it must be employed ten times to simulate one exam. The number of successes is counted for this simulated exam. Then, the process is repeated many times to simulate the distribution of the number of successes.

With \( n = 10, p = 0.25 \), and \( Y \) defined as the number of successes, the distribution of \( Y \) is as given at right:

<table>
<thead>
<tr>
<th>( Y )</th>
<th>( P(Y) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.056314</td>
</tr>
<tr>
<td>1</td>
<td>0.187712</td>
</tr>
<tr>
<td>2</td>
<td>0.281568</td>
</tr>
<tr>
<td>3</td>
<td>0.250282</td>
</tr>
<tr>
<td>4</td>
<td>0.145998</td>
</tr>
<tr>
<td>5</td>
<td>0.058399</td>
</tr>
<tr>
<td>6</td>
<td>0.016222</td>
</tr>
<tr>
<td>7</td>
<td>0.003090</td>
</tr>
<tr>
<td>8</td>
<td>0.000386</td>
</tr>
<tr>
<td>9</td>
<td>0.000029</td>
</tr>
<tr>
<td>10</td>
<td>0.000001</td>
</tr>
</tbody>
</table>

The chance of answering at least 60%, or 6 questions, correctly is the sum of the last five rows of the table, which is 0.0197. The student has a very slight chance of guessing his or her way to a passing grade.

Now, the chance of a correct answer on any one question is \( \frac{1}{3} \).

The probability distribution for the number of correct answers out of ten questions is given by:
From this distribution, the probability of getting 6 or more answers correct is 0.0732, a little larger than in the case of four choices, but not much. The student still has a poor chance of guessing to a passing grade.

d. The distribution of a number of correct answers \( Y \) for \( n = 5 \) and \( p = 0.25 \) is shown below.

\[
\begin{array}{c|c}
Y & P(Y) \\
0 & 0.237305 \\
1 & 0.395508 \\
2 & 0.263672 \\
3 & 0.087891 \\
4 & 0.014648 \\
5 & 0.000977 \\
\end{array}
\]

Now, getting at least 60% correct is the same as getting 3 or more correct answers, which has probability 0.1035. The chance of obtaining a passing grade by guessing is much greater on the short exam than on the long one.

4. a. You are interested in one particular coupon out of five. Thus, your probability of success on any one purchase is \( \frac{1}{5} = 0.2 \), assuming the coupons are randomly distributed in equal numbers. You could sample random digits, counting the number of digits until the first 0 or 1 appears. This process should be repeated many times in order to generate a distribution of values for the number of coupons necessary to achieve success.

\[
P(\text{purchase 4 or fewer boxes}) = 1 - P(\text{the coupon of interest is not in any of the first four boxes}) = 1 - (0.8)^4 = 1 - 0.4096 = 0.5904.
\]

b. \( E(\text{number of boxes purchased}) = \frac{10 \cdot 0.2}{0.2} = 5 \).

c. \( E(\text{number of boxes purchased on 1st stage}) + E(\text{number of boxes purchased on 2nd stage}) = \frac{1}{0.4} + \frac{10}{0.2} = 2.5 + 5 = 7.5 \).
Teacher Resources
1. A quality-improvement plan for an aircraft-repair facility required information on the number of defects seen on repaired pumps coming from the hydraulics shop. A random sample of 20 pumps from the shop showed the following counts on the number of defects per pump:

63 40 27 31 00 43 22 65 07 21

The random variable of interest for planning future inspections is \( Y \), the number of defects per pump.

a. Use the data from the sample of 20 pumps to approximate a probability distribution for the random variable \( Y \).

b. Find the expected value of \( Y \) by using the approximate distribution for \( Y \) found in part a.

c. Find the standard deviation of \( Y \) by using the approximate distribution found in part a.

2. Each defect found in a pump as described in Problem 1—usually a faulty gasket—costs $50 to repair.

a. Find the estimated expected repair cost per pump.

b. Find the estimated standard deviation of repair costs per pump.

c. Suppose $200 per pump is budgeted for repairing these defects. Will this amount be exceeded often? Explain your reasoning.

3. The table below shows the frequency, in millions, for ages of cars in use in the United States for 1980 and 1994. For example, in 1980 there were 52.3 million cars under 5 years old in use across the country.
<table>
<thead>
<tr>
<th>Age in years</th>
<th>1980</th>
<th>1994</th>
</tr>
</thead>
<tbody>
<tr>
<td>5 or less</td>
<td>52.3</td>
<td>45.4</td>
</tr>
<tr>
<td>6 - 8</td>
<td>25.2</td>
<td>27.7</td>
</tr>
<tr>
<td>9 - 11</td>
<td>14.6</td>
<td>25.1</td>
</tr>
<tr>
<td>12 or more</td>
<td>12.5</td>
<td>31.4</td>
</tr>
</tbody>
</table>

Source: Statistical Abstract of the United States, 1996

a. Construct plots of each of these distributions. How do the two distributions differ?

b. Approximate the mean of each distribution.

c. If you could obtain more detailed information on the ages of cars in use, do you think the means would increase, compared to your approximations? Explain your reasoning.

d. Approximate the medians for each distribution. How do they compare?

e. Explain which distribution must have the larger standard deviation without doing the calculations.

4. A merchant stocks a certain perishable item. On any given day, she will have a demand for either two, three, or four of these items with probabilities 0.1, 0.4, and 0.5, respectively. She buys each item for $1.00 and sells each item for $1.20. Any items left over at the end of the day represent a total loss. How many items should the merchant stock in order to maximize her expected daily profit?
1. A quality-improvement plan in an aircraft repair facility calls for sampling a number of hydraulic pumps that have been rebuilt and counting the number of defects that are still present. These defects are usually the result of a faulty gasket. The following table shows the approximate probability distribution of $Y$, the number of defects observed per pump.

<table>
<thead>
<tr>
<th>$Y$</th>
<th>$P(Y)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.20</td>
</tr>
<tr>
<td>1</td>
<td>0.10</td>
</tr>
<tr>
<td>2</td>
<td>0.20</td>
</tr>
<tr>
<td>3</td>
<td>0.15</td>
</tr>
<tr>
<td>4</td>
<td>0.10</td>
</tr>
<tr>
<td>5</td>
<td>0.05</td>
</tr>
<tr>
<td>6</td>
<td>0.10</td>
</tr>
<tr>
<td>7</td>
<td>0.10</td>
</tr>
</tbody>
</table>

a. The next group of aircraft coming in to be repaired will have a total of 100 pumps that need rebuilding. Find the expected value and the standard deviation of the mean number of defects per rebuilt pump that can be anticipated in this sample of 100 pumps.

b. The quality improvement plan set a goal of no more than 230 defects to be observed on the 100 rebuilt pumps. Would the facility be likely to reach that goal if it is still operating under the distribution of defects per pump shown above? Explain your reasoning.
2. The table below shows the frequency distribution of the ages of drivers under the age of 25, called “young adults.”

<table>
<thead>
<tr>
<th>Age</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>16</td>
<td>1,470,521</td>
</tr>
<tr>
<td>17</td>
<td>2,200,842</td>
</tr>
<tr>
<td>18</td>
<td>2,493,137</td>
</tr>
<tr>
<td>19</td>
<td>2,727,972</td>
</tr>
<tr>
<td>20</td>
<td>2,836,091</td>
</tr>
<tr>
<td>21</td>
<td>2,921,521</td>
</tr>
<tr>
<td>22</td>
<td>3,130,025</td>
</tr>
<tr>
<td>23</td>
<td>3,482,642</td>
</tr>
<tr>
<td>24</td>
<td>3,608,980</td>
</tr>
</tbody>
</table>


a. If \( Y \) denotes the age of a driver randomly selected from the population of young-adult drivers, construct a probability distribution for \( Y \).

b. A polling company is to randomly sample 400 young-adult drivers. Describe the distribution of the possible values of the mean age of the drivers in this sample.

c. Suppose a sample of 400 young-adult drivers produced a mean age of 21 years. What questions would this raise in your mind and why?

3. In a recent poll of 1000 Americans, it was found that 81% expressed belief in the existence of heaven (Time, March 24, 1997).

a. What is the approximate margin of error for this poll result?

b. Explain what is meant by the “margin of error.”
According to the U.S. Bureau of the Census, about 15% of Americans have no health insurance. A government agency is designing a sampling plan to interview Americans about their concerns. The plan calls for a sample size of 500.

a. If 500 Americans are randomly sampled, describe the probability distribution of the potential values of the sample proportion of those having no health insurance.

b. The agency would like to locate 100 Americans with no health insurance for further study. Is the sample of 500 likely to produce 100 with no health insurance? Explain your reasoning.
1. According to the U.S. Bureau of the Census, 15% of Americans are without health insurance.

a. Six new patients come through the emergency room of a hospital in one morning. Find the probability distribution of the number of new patients without health insurance. Make a rough plot of the distribution and comment on its shape. What assumption is necessary for your answer to be valid?

b. Fifty new patients come through the emergency room of a hospital in one day. Find the expected value and standard deviation of the number of patients without health insurance among the 50. What assumption is necessary for your answer to be valid?

c. Among the 50 new patients coming through the emergency room, what is the approximate probability that more than ten of them will be without health insurance? A rough approximation will be sufficient.

2. A quality-control plan in a plant manufacturing automobile batteries calls for the random selection and weighing of one battery every hour. Weight is an important quality measurement, because the quality of a battery depends on the amount of lead in its plates. According to the manufacturing specifications, these batteries should have approximately normally distributed weights with a mean of 68 pounds and a standard deviation of 0.5 pound.

a. If the weight of a sampled battery lies more than 3 standard deviations from the specified mean, an “out-of-control” signal is sent to the engineer in charge and a series of further checks on the production process are begun. Find the probability that no out-of-control signals are sent in one 8-hour shift if the process is operating according to specifications.
b. Repeat part a with the change that out-of-control limits are now only 2 standard deviations from the specified mean.

3. Review the quality-control scenario of Problem 2 and assume the out-of-control limits are set at 3 standard deviations.

   a. How many hours can the plant manager expect to wait until the first out-of-control signal appears, assuming that the plant is operating according to specifications?

   b. Find the standard deviation of the number of batteries that must be checked until the first out-of-control signal sounds.

   c. Is there a good chance that the plant may see no out-of-control signals in 600 hours of operation? Explain your reasoning.

4. Your local theater is distributing prizes in the large tubs of popcorn it sells during the reprise of the Star Wars® movies. The prizes are small models of Star Wars spacecraft. Three different models are randomly distributed in equal numbers, one model per tub of popcorn. How many tubs of popcorn can you expect to buy in order to obtain a full set of models, that is, at least one of each of the three types available? Explain how you arrived at this answer.

Note: Star Wars is a registered trademark of Lucasfilm, LTD.
1. a. The data on individual pump defect counts must be placed into a relative-frequency table, since probabilities are estimated by relative frequencies. The table could look like this one, although the frequencies do not need to be recorded.

<table>
<thead>
<tr>
<th>Y</th>
<th>Frequency</th>
<th>P(Y)</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>4</td>
<td>0.20</td>
</tr>
<tr>
<td>1</td>
<td>2</td>
<td>0.10</td>
</tr>
<tr>
<td>2</td>
<td>4</td>
<td>0.20</td>
</tr>
<tr>
<td>3</td>
<td>3</td>
<td>0.15</td>
</tr>
<tr>
<td>4</td>
<td>2</td>
<td>0.10</td>
</tr>
<tr>
<td>5</td>
<td>1</td>
<td>0.05</td>
</tr>
<tr>
<td>6</td>
<td>2</td>
<td>0.10</td>
</tr>
<tr>
<td>7</td>
<td>2</td>
<td>0.10</td>
</tr>
</tbody>
</table>

b. \( E(Y) = 2.90 \)

c. \( SD(Y) = 2.28 \)

2. a. Repair cost per pump, denoted by \( C \), is given by \( C = 50Y \).

\[ E(C) = 50E(Y) = 50(2.9) = 145 \]

b. The standard deviation of repair cost per pump is

\[ SD(C) = 50SD(Y) = 50(2.28) = 114 \]

c. The $200 budgeted amount is about one half a standard deviation above the mean. This amount could be exceeded often.

3. a. Shown below are the relative-frequency, or proportion, plots for the data on ages of cars.

The distribution for 1980 is highly skewed toward the older cars, while the distribution for 1994 is much flatter and, in fact, increases slightly as we move toward the oldest age category. This shows that the cars in use in 1994 tend to be older than those in use in 1980, relatively speaking.

b. Using the midpoints of the intervals, 3, 7, and 10, as typical values and 14 as a typical value for the “12 or more” category, the means are 6.25 years and 7.88 years, respectively.

c. More information would probably show that there are cars in use that are much older than 12 years. In other words, the 14 used above might be too small. The means should increase with more information on the older cars becoming available.

d. The median for the 1980 ages is about 5 and the median for the 1994 ages is a little less than 7. The cars in use in 1994 appear to be older than the ones in use in 1980, on the whole.

The easiest way to see the locations of the medians is to turn the frequency data into relative-frequency data, or estimated probabilities. The result is shown below:
<table>
<thead>
<tr>
<th>Age</th>
<th>( P(Y)'80 )</th>
<th>( P(Y)'94 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>5 or less</td>
<td>0.5000</td>
<td>0.3503</td>
</tr>
<tr>
<td>6–8</td>
<td>0.2409</td>
<td>0.2137</td>
</tr>
<tr>
<td>9–11</td>
<td>0.1396</td>
<td>0.1937</td>
</tr>
<tr>
<td>12 or more</td>
<td>0.1195</td>
<td>0.2423</td>
</tr>
</tbody>
</table>

e. The 1994 data have the more even frequencies in the categories and, hence, the greater spread. The 1980 data has a high percent of values concentrated in the 5-or-under category and relatively low percents in the two oldest categories.

4. The choice is to be made among two, three or four items as the number to stock. The merchant makes a profit of $0.20 on each item sold, but loses $1.00 if an item is not sold. Let \( G \) denote the merchant’s gain. Consider the three choices separately.

Stock two: In this case, the two are sure to sell because there is a probability of 1 that at least two customers will show up. \( G \) has the distribution

<table>
<thead>
<tr>
<th>( G )</th>
<th>( P(G) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.40</td>
<td>1</td>
</tr>
</tbody>
</table>

From this distribution, \( E(G) = 0.40 \).

Stock three: There is a demand for at least three items with probability 0.9. If she stocks three and sells only two, she loses $0.60. \( G \) has the distribution

<table>
<thead>
<tr>
<th>( G )</th>
<th>( P(G) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>-0.60</td>
<td>0.1</td>
</tr>
<tr>
<td>0.60</td>
<td>0.9</td>
</tr>
</tbody>
</table>

From this distribution, \( E(G) = 0.48 \).

Stock four: In this case the distribution of \( G \) is

<table>
<thead>
<tr>
<th>( G )</th>
<th>( P(G) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>-1.60</td>
<td>0.1</td>
</tr>
<tr>
<td>-0.40</td>
<td>0.4</td>
</tr>
<tr>
<td>0.80</td>
<td>0.5</td>
</tr>
</tbody>
</table>

For this distribution, \( E(G) = 0.08 \).

The choice that maximizes expected gain is to stock three items each day.
1. a. From the distribution of the number of defects per pump given in the problem,
\[ E(Y) = 2.90 \text{ and } SD(Y) = 2.28. \]
The distribution of the mean number of defects per pump in a random sample of 100 will center at \( E(Y) = 2.90 \) and will have standard deviation given by \( SD(Y) = \frac{SD(Y)}{\sqrt{n}} = \frac{2.28}{\sqrt{10}} = 0.228. \)

b. For the total defects to drop below 230, the mean number of defects per pump must drop below 2.30. This value of 2.30 is about 2.6 standard deviations below the center of the sampling distribution of means. Since the sampling distribution of means should be approximately a normal distribution, a result this small would be very unusual. The facility is not likely to reach this goal of 230 defects unless it improves its basic distribution of defects per pump.

c. A mean age of 21 is about 3.5 standard deviations above the center of the sampling distribution for means from random samples. Such a result would be very unusual, since the sampling was done randomly. It might be that the sample was not random or that the population being sampled actually included some drivers older than young adult.

2. a. Turning the frequencies into relative frequencies, or proportions, produces the following:

<table>
<thead>
<tr>
<th>( Y )</th>
<th>( P(Y) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>16</td>
<td>0.0591</td>
</tr>
<tr>
<td>17</td>
<td>0.0885</td>
</tr>
<tr>
<td>18</td>
<td>0.1002</td>
</tr>
<tr>
<td>19</td>
<td>0.1097</td>
</tr>
<tr>
<td>20</td>
<td>0.1140</td>
</tr>
<tr>
<td>21</td>
<td>0.1174</td>
</tr>
<tr>
<td>22</td>
<td>0.1258</td>
</tr>
<tr>
<td>23</td>
<td>0.1400</td>
</tr>
<tr>
<td>24</td>
<td>0.1451</td>
</tr>
</tbody>
</table>

b. From the above distribution, \( E(Y) = 20.56 \) years and \( SD(Y) = 2.48 \) years. The sampling distribution of the sample mean for a sample of size 400 will be approximately normal in distribution with mean of 20.56 and standard deviation of \( \frac{2.48}{\sqrt{20}} = 0.124. \)

c. The approximate margin of error is 2 estimated standard deviations of the sample proportion. This turns out to be
\[ 2SD(\text{proportion}) = 2 \sqrt{\frac{(0.81)(0.19)}{1000}} = 0.0248. \]

b. The margin of error is a distance between the sample proportion and the population proportion that will be bettered 95% of the time in repeated use of the technique. With probability 0.95, the sample proportion of those who believe in heaven will not differ from the population proportion who believe the same by more than 0.0248, in samples of size 1000. You can be confident that the 0.81 observed in the sample is reasonably close to the true proportion of those who believe in the existence of heaven.

d. The potential values of the sample proportion of those without health insurance, in a random sample of size 500, would have a normal distribution with mean 0.15 and standard deviation given by
\[ SD(\text{proportion}) = \sqrt{\frac{(0.15)(0.85)}{500}} = 0.016. \]

b. To find at least 100 Americans with no health insurance in a sample of 500, the sample proportion of those without health insurance would have to be at least 0.2. This value is over three standard deviations above the mean of the sampling distribution. Thus, it is very unlikely that the agency will obtain 100 people without health insurance in a random sample of only 500.
1. a. The probability distribution for a binomial random variable with $n = 6$ and $p = 0.15$ is shown below in a table and in a plot. The distribution is highly skewed toward more patients because the small value of $p$ concentrates most of the outcomes around 0 and 1. Note that 0 does not have the highest probability of occurrence, even though 0.15 is closer to 0 than it is to 1.

The six patients would have to behave as a random selection from a population of patients with 15% having no health insurance in order for these probabilities to be valid. If, for example, the hospital was in a section of a city where more than 15% are without health insurance, then this would not be the correct distribution.

<table>
<thead>
<tr>
<th>$Y$</th>
<th>$P(Y)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.377149</td>
</tr>
<tr>
<td>1</td>
<td>0.399335</td>
</tr>
<tr>
<td>2</td>
<td>0.176177</td>
</tr>
<tr>
<td>3</td>
<td>0.041453</td>
</tr>
<tr>
<td>4</td>
<td>0.005486</td>
</tr>
<tr>
<td>5</td>
<td>0.000387</td>
</tr>
<tr>
<td>6</td>
<td>0.000011</td>
</tr>
</tbody>
</table>

b. For a binomial distribution with $n = 50$ and $p = 0.15$, $E(Y) = np = 50(0.15) = 7.5$ and $SD(Y) = \sqrt{np(1-p)} = \sqrt{50(0.15)(0.85)} = 2.52$.

Again, the sample of 50 must be a random sample from a population with 15% having no health insurance in order for these quantities to be valid.

c. Possible answer: A quick answer is that, since 10 is about 1 standard deviation above the mean, the chance of seeing more than 10 patients without health insurance is about 0.16. The binomial distribution for $n = 50$ will be approximately normal even though $p$ is quite small. The exact answer for the sum of the binomial probabilities greater than 10 is 0.120.

2. a. The measured weight of a battery is out of control if it is more than 3 standard deviations from the mean, under a normal distribution. This probability is 0.003. The chance of not being out of control is 0.997. For independent measurements, the chance of not being out of control 8 times in a row is $(0.997)^8 = 0.976$.

b. If the control limits shift to 2 standard deviations from the mean, then the probability of being within them is 0.95. The probability of being within them 8 times in a row is $(0.95)^8 = 0.663$.

Note that this change in the control limits makes a big difference in the chances of being out of control at least once during the day.

c. If $Y$ is the number of hours or the number of sampled batteries until an out-of-control signal flashes, then $Y$ has a geometric distribution and $E(Y) = \frac{1}{p} = \frac{1}{0.003} = 333.3$.

$SD(Y) = \sqrt{\frac{1-p}{p}} = \sqrt{\frac{0.997}{0.003}} = 332.8$

Note that the expected value and standard deviation are nearly equal when $p$ is small.

c. The quick answer is that, since 600 is only about 1 standard deviation above the mean, it must have a good chance of occurring. The exact answer is $(0.997)^{600} = 0.165$. This is most easily calculated by using logarithms.

3. a. Think of this as a series of three waiting-time problems. When starting out, you will want whichever prize comes up, so the probability of success is 1. At the second stage, you are looking for one of the two remaining prizes, so your
probability of success is \( \frac{2}{3} \). After you get one of the two, you are looking for the single prize that is not in your collection. Thus, at the third stage your probability of success is \( \frac{1}{3} \). Since all the selections of prizes are independent, your expected number of tubs of popcorn is the sum of the expected waiting times for the three stages, namely, 

\[ 1 + \frac{3}{2} + \frac{3}{1} = 5.5. \]

You can expect to buy 5.5 tubs of popcorn to complete your collection.
Data-Driven Mathematics

Procedures for Using the TI-83

I. Clear menus
ENTER will execute any command or selection.
Before beginning a new problem, previous instructions or data should be cleared. Press ENTER after each step below.

1. To clear the function menu, Y=, place the cursor anyplace in each expression, CLEAR
2. To clear the list menu, 2nd MEM ClrAllLists
3. To clear the draw menu, 2nd DRAW ClrDraw
4. To turn off any statistics plots, 2nd STATPLOT PlotsOff
5. To remove user created lists from the Editor, STAT SetUpEditor

II. Basic information
1. A rule is active if there is a dark rectangle over the option. See Figure 1.

![Figure 1](image)

On the screen above, Y1 and Plot1 are active; Y2 is not. You may toggle Y1 or Y2 from active to inactive by putting the cursor over the = and pressing ENTER. Arrow up to Plot1 and press ENTER to turn it off; arrow right to Plot2 and press ENTER to turn it on, etc.

2. The Home Screen (Figure 2) is available when the blinking cursor is on the left as in the diagram below. There may be other writing on the screen. To get to the Home Screen, press 2nd QUIT. You may also clear the screen completely by pressing CLEAR.

![Figure 2](image)

3. The variable x is accessed by the X, T, Θ, n key.

4. Replay option: 2nd ENTER allows you to back up to an earlier command. Repeated use of 2nd ENTER continues to replay earlier commands.

5. Under MATH, the MATH menu has options for fractions to decimals and decimals to fractions, for taking nth roots, and for other mathematical operations. NUM contains the absolute value function as well as other numerical operations. (Figure 3)

![Figure 3](image)

III. The STAT Menus

1. There are three basic menus under the STAT key: EDIT, CALC, and TESTS. Data are entered and modified in the EDIT mode; all numerical calculations are made in the CALC mode; statistical tests are run in the TEST mode.

2. Lists and Data Entry
Data is entered and stored in Lists (Figure 4). Data will remain in a list until the list is cleared. Data can be cleared using CLEAR Lists or (List name), or by placing the cursor over the List heading and selecting CLEAR ENTER. To enter data, select STAT EDIT and with the arrow keys move the cursor to the list you want to use.
Type in a numerical value and press ENTER. Note that the bottom of the screen indicates the List you are in and the list element you have highlighted. 275 is the first entry in L1. (It is sometimes easier to enter a complete list before beginning another.)

L1  L2  L3
[275  67  190]
5311  144  120
114  64  238
2838  111  153
15  90  179
332  68  207
3828  94  153

L1 (1) = 275

FIGURE 4

For data with varying frequencies, one list can be used for the data, and a second for the frequency of the data. In Figure 5 below, the L5(7) can be used to indicate that the seventh element in list 5 is 4, and that 25 is a value that occurs 4 times.

L4  L5  L6
[55  1  
50  3  
45  6  
40  14 
35  12 
30  9  
25  4 

L5 (7) = 4

FIGURE 5

3. Naming Lists

Six lists are supplied to begin with. L1, L2, L3, L4, L5, and L6 can be accessed also as 2nd L1. Other lists can be named using words as follows. Put the cursor at the top of one of the lists. Press 2nd INS and the screen will look like that in Figure 6.

FIGURE 6

The alpha key is on, so type in the name (up to five characters) and press ENTER. (Figure 7)

FIGURE 7

Then enter the data as before. (If you do not press ENTER, the cursor will remain at the top and the screen will say "error: data type." ) The newly named list and the data will remain until you go to Memory and delete the list from the memory. To access the list for later use, press 2nd LIST and use the arrow key to locate the list you want under the NAMES menu. You can accelerate the process by typing ALPHA P (for price). (Figure 8) Remember, to delete all but the standard set of lists from the editor, use SetUp Editor from the STAT menu.

FIGURE 8
4. Graphing Statistical Data

General Comments
- Any graphing uses the GRAPH key.
- Any function entered in Y1 will be graphed if it is active. The graph will be visible only if a suitable viewing window is selected.
- The appropriate x- and y-scales can be selected in WINDOW. Be sure to select a scale that is suitable to the range of the variables.

Statistical Graphs
To make a statistical plot, select 2nd Y= for the STAT PLOT option. It is possible to make three plots concurrently if the viewing windows are identical. In Figure 9, Plots 2 and 3 are off, Plot 1 is a scatter plot of the data (Costs, Seats), Plot 2 is a scatter plot of (L3, L4), and Plot 3 is a box plot of the data in L3.

Activate one of the plots by selecting that plot and pressing ENTER.

Choose ON, then use the arrow keys to select the type of plot (scatter, line, histogram, box plot with outliers, box plot, or normal probability plot). (In a line plot, the points are connected by segments in the order in which they are entered. It is best used with data over time.)

Choose the lists you wish to use for the plot. In the window below, a scatter plot has been selected with the x-coordinate data from COSTS, and the y-coordinate data from SEATS. (Figure 10) (When pasting in list names, press 2nd LIST, press ENTER to activate the name, and press ENTER again to locate the name in that position.)

5. Statistical Calculations
One-variable calculations such as mean, median, maximum value of the list, standard deviation, and quartiles can be found by selecting STAT CALC 1-Var Stats followed by the list in which you are interested. Use the arrow to continue reading the statistics. (Figures 11, 12, 13)
Calculations of numerical statistics for bivariate data can be made by selecting two variable statistics. Specific lists must be selected after choosing the 2-Var Stats option. (Figure 14)

Individual statistics for one- or two-data sets can be obtained by selecting VARS Statistics, but you must first have calculated either 1-Var or 2-Var Statistics. (Figure 15)

6. Fitting Lines and Drawing Their Graphs

Calculations for fitting lines can be made by selecting the appropriate model under STAT CALC: Med-Med gives the median fit regression, LinReg the least-squares linear regression, and so on. (Note the only difference between LinReg (ax+b) and LinReg (a+bx) is the assignment of the letters a and b.) Be sure to specify the appropriate lists for x and y. (Figure 16)

To graph a regression line on a scatter plot, follow the steps below:
- Enter your data into the Lists.
- Select an appropriate viewing window and set up the plot of the data as above.
- Select a regression line followed by the lists for x and y, VARS Y-VARS Function (Figures 17, 18) and the Y, you want to use for the equation, followed by ENTER.
The result will be the regression equation pasted into the function Y1. Press GRAPH and both the scatter plot and the regression line will appear in the viewing window. (Figures 19, 20)

There are two cursors that can be used in the graphing screen.

TRACe activates a cursor that moves along either the data (Figure 21) or the function entered in the Y-variable menu (Figure 22). The coordinates of the point located by the cursor are given at the bottom of the screen.

Pressing GRAPH returns the screen to the original plot. The up arrow key activates a cross cursor that can be moved freely about the screen using the arrow keys. See Figure 23.

Exact values can be obtained from the plot by selecting 2nd CALC Value. Select 2nd CALC Value ENTER. Type in the value of x you would like to use, and the exact ordered pair will appear on the screen with the cursor located at that point on the line. (Figure 24)
VII. Random Numbers

To generate random numbers, press MATH and PRB. This will give you access to a random number function, \texttt{rand}, that will generate random numbers between 0 and 1 or \texttt{randInt()} that will generate random numbers from a beginning integer to an ending integer for a specified set of numbers. (Figure 35) In Figure 36, five random numbers from 1 to 6 were generated.

\begin{center}
\begin{tabular}{cc}
\textbf{MATH} & \textbf{NUM} \\
\textbf{CPX} & \textbf{PRB} \\
1: & \texttt{rand} \\
2: & \texttt{nPr} \\
3: & \texttt{nCr} \\
4: & ! \\
5: & \texttt{randInt()} \\
6: & \texttt{randNorm()} \\
7: & \texttt{randBin()} \\
\end{tabular}
\end{center}

\textbf{FIGURE 35}

\begin{quote}
\texttt{randInt(1, 6, 5, )} \\
\texttt{(2 4 2 5 3)}
\end{quote}

\textbf{FIGURE 36}

Pressing ENTER will generate a second set of random numbers.
Data-Driven Mathematics is a series of modules written by teachers and statisticians that focuses on the use of real data and statistics to motivate traditional mathematics topics. This chart suggests which modules could be used to supplement specific middle-school and high-school mathematics courses.

INTRODUCTORY ALGEBRA

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PROBABILITY THROUGH DATA

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EXPLORING PROJECTS

EXPLORING LINEAR RELATIONS

EXPLORING SYSTEMS OF INEQUALITIES

ALGEBRA

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PROBABILITY MODELS

EXPLORING REGRESSION

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