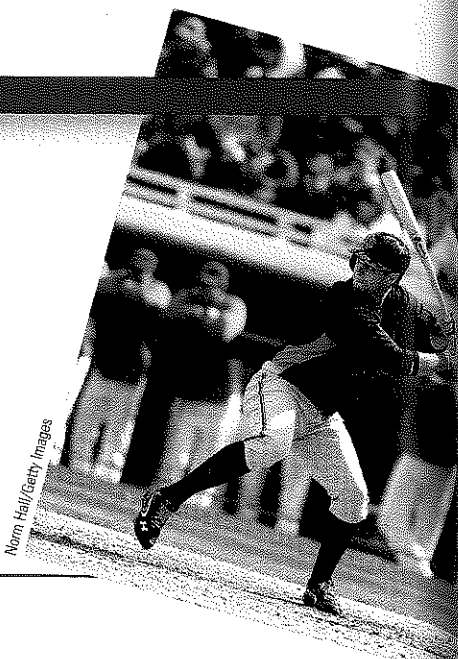


# Module 7

## Statistical Reasoning in Everyday Life

### Module Learning Objectives

- 7-1** Describe the three measures of central tendency, and discuss the relative usefulness of the two measures of variation.
- 7-2** Explain how we know whether an observed difference can be generalized to other populations.



Norm Hall/Getty Images

In descriptive, correlational, and experimental research, statistics are tools that help us see and interpret what the unaided eye might miss. Sometimes the unaided eye misses badly. Researchers invited 5522 Americans to estimate the percentage of wealth possessed by the richest 20 percent in their country (Norton & Ariely, 2011). Their average person's guess—58 percent—"dramatically underestimated" the actual wealth inequality. (The wealthiest 20 percent possess 84 percent of the wealth.)

### The Need for Statistics

Accurate statistical understanding benefits everyone. To be an educated person today is to be able to apply simple statistical principles to everyday reasoning. One needn't memorize complicated formulas to think more clearly and critically about data.

Off-the-top-of-the-head estimates often misread reality and then mislead the public. Someone throws out a big, round number. Others echo it, and before long the big, round number becomes public misinformation. A few examples:

- *Ten percent of people are lesbians or gay men.* Or is it 2 to 3 percent, as suggested by various national surveys (Module 53)?
- *We ordinarily use but 10 percent of our brain.* Or is it closer to 100 percent (Module 12)?
- *The human brain has 100 billion nerve cells.* Or is it more like 40 billion, as suggested by extrapolation from sample counts (Module 10)?

*The point to remember:* Doubt big, round, undocumented numbers.

Statistical illiteracy also feeds needless health scares (Gigerenzer et al., 2008, 2009, 2010). In the 1990s, the British press reported a study showing



"Figures can be misleading—so I've written a song which I think expresses the real story of the firm's performance this quarter."

#### FYI

Asked about the *ideal* wealth distribution in America, Democrats and Republicans were surprisingly similar. In the Democrats' ideal world, the richest 20 percent would possess 30 percent of the wealth. The Republicans' ideal world was similar, with the richest 20 percent possessing 35 percent of the wealth. (Norton & Ariely, 2011).

#### AP® Exam Tip

Do math and statistics scare you? Take a couple of deep breaths and relax before continuing. You will not be asked to do difficult computations on the AP® exam. Nothing will be beyond the scope of simple mental math. You need to focus on the concepts. Why do these statistics exist? How can they help us understand the real world?

that women taking a particular contraceptive pill had a 100 percent increased risk of blood clots that could produce strokes. This caused thousands of women to stop taking the pill, leading to a wave of unwanted pregnancies and an estimated 13,000 additional abortions (which also are associated with increased blood clot risk). And what did the study find? A 100 percent increased risk, indeed—but only from 1 in 7000 women to 2 in 7000 women. Such false alarms underscore the need to teach statistical reasoning and to present statistical information more transparently.

### Descriptive Statistics

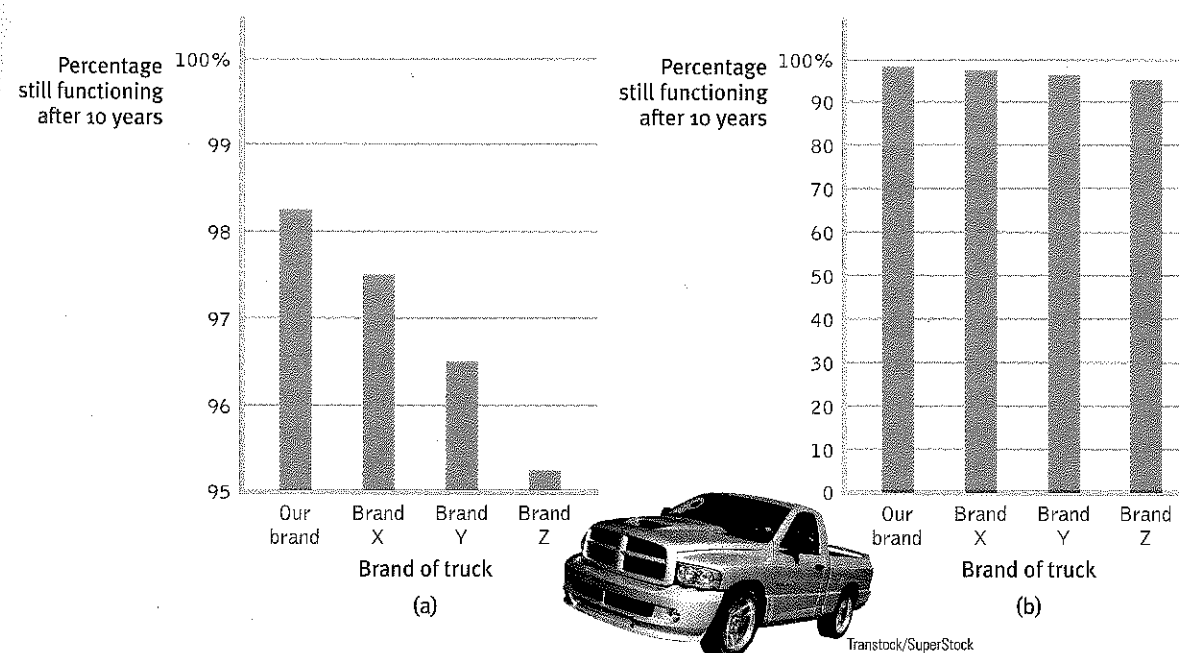
- 7-1** How do we describe data using three measures of central tendency, and what is the relative usefulness of the two measures of variation?

Once researchers have gathered their data, they may use **descriptive statistics** to organize that data meaningfully. One way to do this is to convert the data into a simple *bar graph*, called a **histogram**, as in **FIGURE 7.1**, which displays a distribution of different brands of trucks still on the road after a decade. When reading statistical graphs such as this, take care. It's easy to design a graph to make a difference look big (Figure 7.1a) or small (Figure 7.1b). The secret lies in how you label the vertical scale (the *y-axis*).

*The point to remember:* Think smart. When viewing figures in magazines and on television, read the scale labels and note their range.

### Measures of Central Tendency

The next step is to summarize the data using some *measure of central tendency*, a single score that represents a whole set of scores. The simplest measure is the **mode**, the most frequently occurring score or scores. The most commonly reported is the **mean**, or arithmetic average—the total sum of all the scores divided by the number of scores. On a divided highway, the median is the middle. So, too, with data: The **median** is the midpoint—the 50th percentile. If you arrange all the scores in order from the highest to the lowest, half will be above the median and half will be below it. In a symmetrical, bell-shaped distribution of scores, the mode, mean, and median scores may be the same or very similar.



**Figure 7.1**

**Read the scale labels** An American truck manufacturer offered graph (a)—with actual brand names included—to suggest the much greater durability of its trucks. Note, however, how the apparent difference shrinks as the vertical scale changes in graph (b).

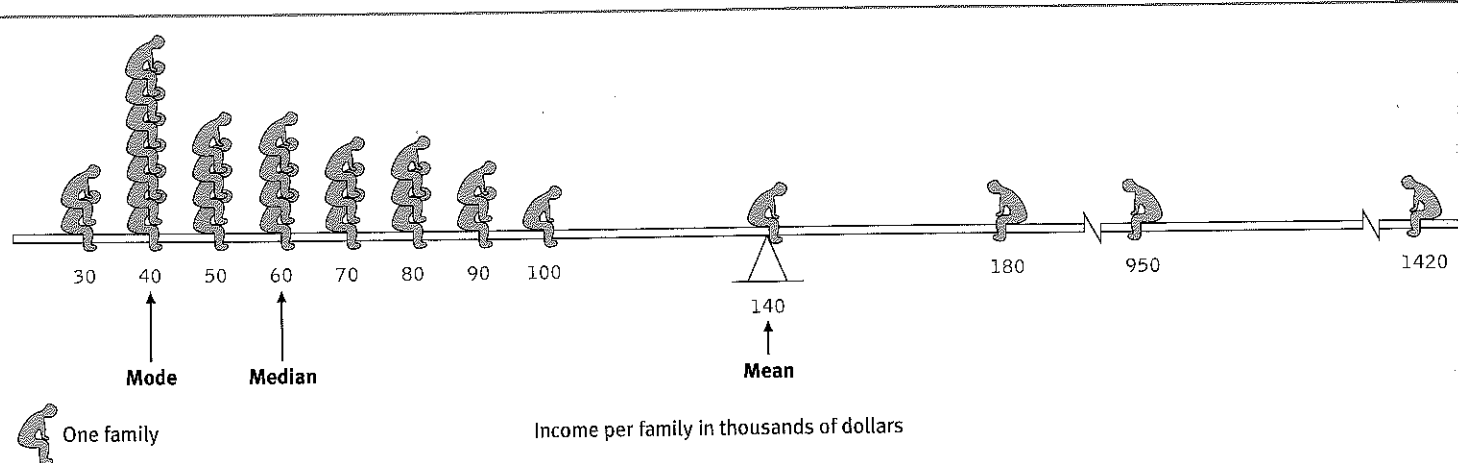


Transtock/SuperStock

Measures of central tendency neatly summarize data. But consider what happens to the mean when a distribution is lopsided, or **skewed**, by a few way-out scores. With income data, for example, the mode, median, and mean often tell very different stories (**FIGURE 7.2**). This happens because the mean is biased by a few extreme scores. When Microsoft co-founder Bill Gates sits down in an intimate café, its average (mean) customer instantly becomes a billionaire. But the customers' median wealth remains unchanged. Understanding this, you can see how a British newspaper could accurately run the headline "Income for 62% Is Below Average" (Waterhouse, 1993). Because the bottom *half* of British income earners receive only a *quarter* of the national income cake, most British people, like most people everywhere, make less than the mean. Mean and median tell different true stories.

*The point to remember:* Always note which measure of central tendency is reported. If it is a mean, consider whether a few atypical scores could be distorting it.

**FYI**  
The average person has one ovary and one testicle.



**Figure 7.2**  
A skewed distribution This graphic representation of the distribution of a village's incomes illustrates the three measures of central tendency—mode, median, and mean. Note how just a few high incomes make the mean—the fulcrum point that balances the incomes above and below—deceptively high.

**Measures of Variation**

Knowing the value of an appropriate measure of central tendency can tell us a great deal. But the single number omits other information. It helps to know something about the amount of *variation* in the data—how similar or diverse the scores are. Averages derived from scores with low variability are more reliable than averages based on scores with high variability. Consider a basketball player who scored between 13 and 17 points in each of her first 10 games in a season. Knowing this, we would be more confident that she would score near 15 points in her next game than if her scores had varied from 5 to 25 points.

The **range** of scores—the gap between the lowest and highest scores—provides only a crude estimate of variation. A couple of extreme scores in an otherwise uniform group, such as the \$950,000 and \$1,420,000 incomes in Figure 7.2, will create a deceptively large range.

The more useful standard for measuring how much scores deviate from one another is the **standard deviation**. It better gauges whether scores are packed together or dispersed, because it uses information from each score (**TABLE 7.1**). The computation assembles information about how much individual scores differ from the mean. If your high school serves a community where most families have similar incomes, family income data will have a relatively small standard deviation compared with the more diverse community population outside your school.

You can grasp the meaning of the standard deviation if you consider how scores tend to be distributed in nature. Large numbers of data—heights, weights, intelligence scores, grades (though not incomes)—often form a symmetrical, *bell-shaped* distribution.

**skewed distribution** a representation of scores that lack symmetry around their average value.

**range** the difference between the highest and lowest scores in a distribution.

**standard deviation** a computed measure of how much scores vary around the mean score.

**Table 7.1** Standard Deviation Is Much More Informative Than Mean Alone

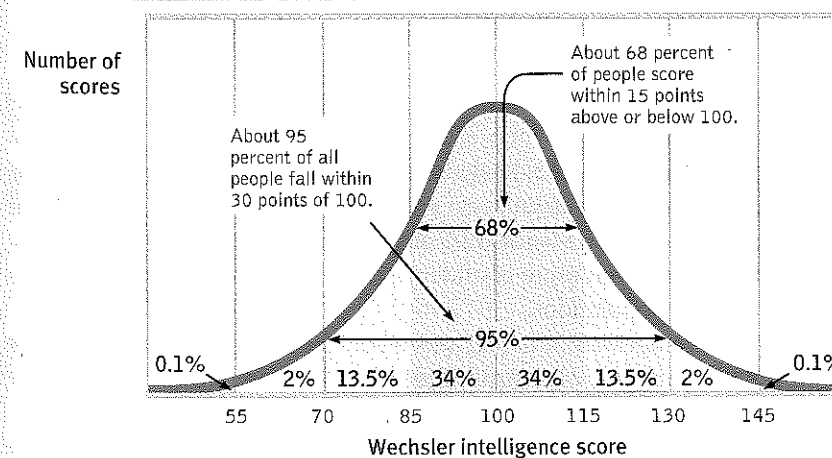
Note that the test scores in Class A and Class B have the same mean (80), but very different standard deviations, which tell us more about how the students in each class are really faring.

Test Scores in Class A			Test Scores in Class B		
Score	Deviation from the Mean	Squared Deviation	Score	Deviation from the Mean	Squared Deviation
72	-8	64	60	-20	400
74	-6	36	60	-20	400
77	-3	9	70	-10	100
79	-1	1	70	-10	100
82	+2	4	90	+10	100
84	+4	16	90	+10	100
85	+5	25	100	+20	400
87	+7	49	100	+20	400
Total = 640	Sum of (deviations) <sup>2</sup> = 204		Total = 640	Sum of (deviations) <sup>2</sup> = 2000	
Mean = 640 ÷ 8 = 80			Mean = 640 ÷ 8 = 80		
Standard deviation = $\sqrt{\frac{\text{Sum of (deviations)}^2}{\text{Number of scores}}} = \sqrt{\frac{204}{8}} = 5.0$			Standard deviation = $\sqrt{\frac{\text{Sum of (deviations)}^2}{\text{Number of scores}}} = \sqrt{\frac{2000}{8}} = 15.8$		

Most cases fall near the mean, and fewer cases fall near either extreme. This bell-shaped distribution is so typical that we call the curve it forms the **normal curve**.

As **FIGURE 7.3** shows, a useful property of the normal curve is that roughly 68 percent of the cases fall within one standard deviation on either side of the mean. About 95 percent of cases fall within two standard deviations. Thus, as Module 61 notes, about 68 percent of people taking an intelligence test will score within ±15 points of 100. About 95 percent will score within ±30 points.

**normal curve (normal distribution)** a symmetrical, bell-shaped curve that describes the distribution of many types of data; most scores fall near the mean (about 68 percent fall within one standard deviation of it) and fewer and fewer near the extremes.



**Figure 7.3**  
The normal curve Scores on aptitude tests tend to form a normal, or bell-shaped, curve. For example, the most commonly used intelligence test, the Wechsler Adult Intelligence Scale, calls the average score 100.

## Inferential Statistics

**7-2** How do we know whether an observed difference can be generalized to other populations?

Data are “noisy.” The average score in one group (breast-fed babies) could conceivably differ from the average score in another group (bottle-fed babies) not because of any real difference but merely because of chance fluctuations in the people sampled. How confidently, then, can we infer that an observed difference is not just a fluke—a chance result of your sampling? For guidance, we can ask how reliable and significant the differences are. These **inferential statistics** help us determine if results can be generalized to a larger population.

**inferential statistics** numerical data that allow one to generalize—to infer from sample data the probability of something being true of a population.

### When Is an Observed Difference Reliable?

In deciding when it is safe to generalize from a sample, we should keep three principles in mind.

1. **Representative samples are better than biased samples.** As noted in Module 5, the best basis for generalizing is not from the exceptional and memorable cases one finds at the extremes but from a representative sample of cases. Research never randomly samples the whole human population. Thus, it pays to keep in mind what population a study has sampled.
2. **Less-variable observations are more reliable than those that are more variable.** As we noted in the example of the basketball player whose game-to-game points were consistent, an average is more reliable when it comes from scores with low variability.
3. **More cases are better than fewer.** An eager high school senior visits two university campuses, each for a day. At the first, the student randomly attends two classes and discovers both instructors to be witty and engaging. At the next campus, the two sampled instructors seem dull and uninspiring. Returning home, the student (discounting the small sample size of only two instructors at each institution) tells friends about the “great instructors” at the first school, and the “bores” at the second. Again, we know it but we ignore it: *Averages based on many cases are more reliable* (less variable) than averages based on only a few cases.

*The point to remember:* Smart thinkers are not overly impressed by a few anecdotes. Generalizations based on a few unrepresentative cases are unreliable.

### When Is a Difference Significant?

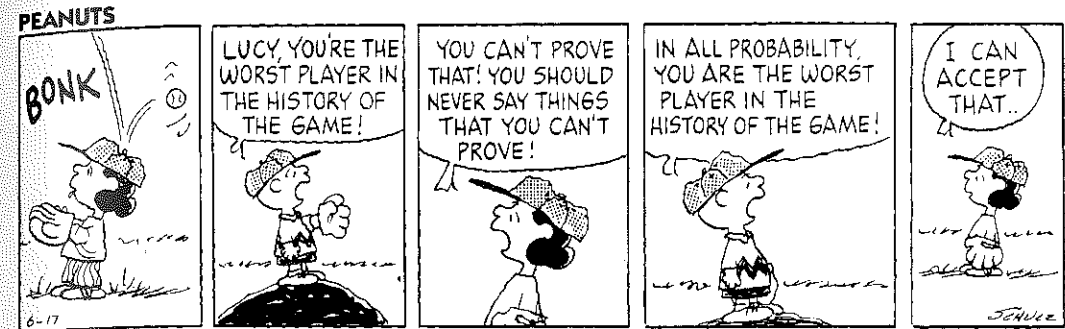
Perhaps you’ve compared men’s and women’s scores on a laboratory test of aggression, and found a gender difference. But individuals differ. How likely is it that the gender difference you found was just a fluke? Statistical testing can estimate the probability of the result occurring by chance.

Here is the underlying logic: When averages from two samples are each reliable measures of their respective populations (as when each is based on many observations that have small variability), then their *difference* is likely to be reliable as well. (Example: The less the variability in women’s and in men’s aggression scores, the more confidence we would have that any observed gender difference is reliable.) And when the difference between the sample averages is *large*, we have even more confidence that the difference between them reflects a real difference in their populations.

In short, when sample averages are reliable, and when the difference between them is relatively large, we say the difference has **statistical significance**. This means that the observed difference is probably not due to chance variation between the samples.

In judging statistical significance, psychologists are conservative. They are like juries who must presume innocence until guilt is proven. For most psychologists, proof beyond a

**statistical significance** a statistical statement of how likely it is that an obtained result occurred by chance.



reasonable doubt means not making much of a finding unless the odds of its occurring by chance, if no real effect exists, are less than 5 percent.

When reading about research, you should remember that, given large enough samples, a difference between them may be “statistically significant” yet have little practical significance. For example, comparisons of intelligence test scores among hundreds of thousands of first-born and later-born individuals indicate a highly significant tendency for first-born individuals to have higher average scores than their later-born siblings (Kristensen & Bjerkedal, 2007; Zajonc & Markus, 1975). But because the scores differ by only one to three points, the difference has little practical importance.

*The point to remember:* Statistical significance indicates the *likelihood* that a result will happen by chance. But this does not say anything about the *importance* of the result.

### AP® Exam Tip

Sometimes a phrase that is frequently used in the media has a more specific meaning when used in psychology. That’s the case with the phrase “statistically significant.” Make sure you know the precise meaning.

## Before You Move On

### ► ASK YOURSELF

Find a graph in a popular magazine ad. How does the advertiser use (or abuse) statistics to make a point?

### ► TEST YOURSELF

Can you solve this puzzle?

The registrar’s office at the University of Michigan has found that usually about 100 students in Arts and Sciences have perfect grades at the end of their first term at the University. However, only about 10 to 15 students graduate with perfect grades. What do you think is the most likely explanation for the fact that there are more perfect grades after one term than at graduation (Jepson et al., 1983)?

*Answers to the Test Yourself questions can be found in Appendix E at the end of the book.*

## Module 7 Review

7-1

How do we describe data using three measures of central tendency, and what is the relative usefulness of the two measures of variation?

- A measure of central tendency is a single score that represents a whole set of scores. Three such measures are the *mode* (the most frequently occurring score), the *mean* (the arithmetic average), and the *median* (the middle score in a group of data).
- Measures of variation tell us how diverse data are. Two measures of variation are the *range* (which describes the gap between the highest and lowest scores) and the *standard deviation* (which states how much scores vary around the mean, or average, score).
- Scores often form a *normal* (or bell-shaped) *curve*.



How do we know whether an observed difference can be generalized to other populations?

- To feel confident about generalizing an observed difference to other populations, we would want to know that
  - the sample studied was representative of the larger population being studied;

- the observations, on average, had low variability;
- the sample consisted of more than a few cases; and
- the observed difference was *statistically significant*.

### Multiple-Choice Questions

- Which of the following is a measure of variation?
  - Range
  - Mean
  - Mode
  - Frequency
  - Median
- Which statistical measure of central tendency is most affected by extreme scores?
  - Mean
  - Median
  - Mode
  - Skew
  - Correlation
- A researcher calculates statistical significance for her study and finds a 5 percent chance that results are due to chance. Which of the following is an accurate interpretation of this finding?
  - This is well beyond the range of statistical significance.
  - This is the minimum result typically considered statistically significant.
  - This is not statistically significant.
  - There is no way to determine statistical significance without replication of the study.
  - Chance or coincidence is unrelated to statistical significance.
- Descriptive statistics \_\_\_\_\_, while inferential statistics \_\_\_\_\_.
  - indicate the significance of the data; summarize the data
  - describe data from experiments; describe data from surveys and case studies
  - are measures of central tendency; are measures of variance
  - determine if data can be generalized to other populations; summarize data
  - summarize data; determine if data can be generalized to other populations
- In a normal distribution, what percentage of the scores in the distribution falls within one standard deviation on either side of the mean?
  - 34 percent
  - 40 percent
  - 50 percent
  - 68 percent
  - 95 percent

### Practice FRQs

- Explain the difference between descriptive and inferential statistics in research.

Answer (2 points)

*1 point:* Descriptive statistics organize and summarize the data collected during research.

*1 point:* Inferential statistics are used to help determine whether results can be generalized to a larger population through the calculation of statistical significance.

- The following data set includes information from survey research in a psychology course regarding how many hours each individual in the class spent preparing for the exam.

Student	Amount of hours reported studying
1	2
2	3
3	6
4	8
5	9
6	9
7	21

Examine the data and respond to the following:

- What is the middle score in this distribution? What term is used to describe the middle score?
- What would be the most useful statistic for measuring the variation of the hours spent studying? Why is this statistic a better measure of variation than the range?

(3 points)