

**STA 2023**

**Module 7**

**Confidence Intervals for  
Proportions**

# Learning Objectives

Upon completing this module, you should be able to:

1. Find and interpret a large-sample confidence interval for a population proportion.
2. Compute the margin of error for the estimate of a population proportion.
3. Describe the relationship between the sample size, confidence interval, and margin of error for a confidence interval for a population proportion.
4. Determine the sample size required for a specified confidence level and margin of error for the estimate of a population proportion.

# What is Population Proportion?

As we have learned from the last module, a **population proportion,  $p$**  is simply the percentage of a population that has a specified attribute. A **population proportion** is also known as **a true proportion**.

# Population Proportion and Sample Proportion

## Population Proportion and Sample Proportion

Consider a population in which each member either has or does not have a specified attribute. Then we use the following notation and terminology.

**Population proportion,  $p$ :** The proportion (percentage) of the entire population that has the specified attribute.

**Sample proportion,  $\hat{p}$ :** The proportion (percentage) of a sample from the population that has the specified attribute.

In short, a **sample proportion** is obtained by dividing **the number of members sampled that have the specified attribute ( $x$ )** by the **total number of members sampled ( $n$ )**.

Sometimes, we refer to “ $x$ ” as the **number of successes** and “ $n-x$ ” as the **number of failures**.

# The Sampling Distribution of the Sample Proportion

In order to make inferences about a population proportion, we must know the **sampling distribution of the sample proportion**.

## The Sampling Distribution of the Sample Proportion

For samples of size  $n$ ,

- the mean of  $\hat{p}$  equals the population proportion:  $\mu_{\hat{p}} = p$ ;
- the standard deviation of  $\hat{p}$  equals the square root of the product of the population proportion and one minus the population proportion divided by the sample size:  $\sigma_{\hat{p}} = \sqrt{p(1 - p)/n}$ ; and
- $\hat{p}$  is approximately normally distributed for large  $n$ .

In short, if  $n$  is large, the possible sample proportions for samples of size  $n$  have approximately a **normal distribution** with **mean  $p$**  and **standard deviation** as above.

# The Sampling Distribution Model of Sample Proportions

- The sampling distribution model of  $\hat{p}$  is centered at  $p$ , with standard deviation  $\sqrt{\frac{pq}{n}}$ , where  $q = 1 - p$ .
- Since we don't know  $p$ , we can't find the true standard deviation (population standard deviation) of the sampling distribution model, so we need to find the standard error:  $SE(\hat{p}) = \sqrt{\frac{\hat{p}\hat{q}}{n}}$

Note that  $\hat{p}$  (read as p hat) is the sample proportion, whereas  $p$  is the true population proportion (true proportion).

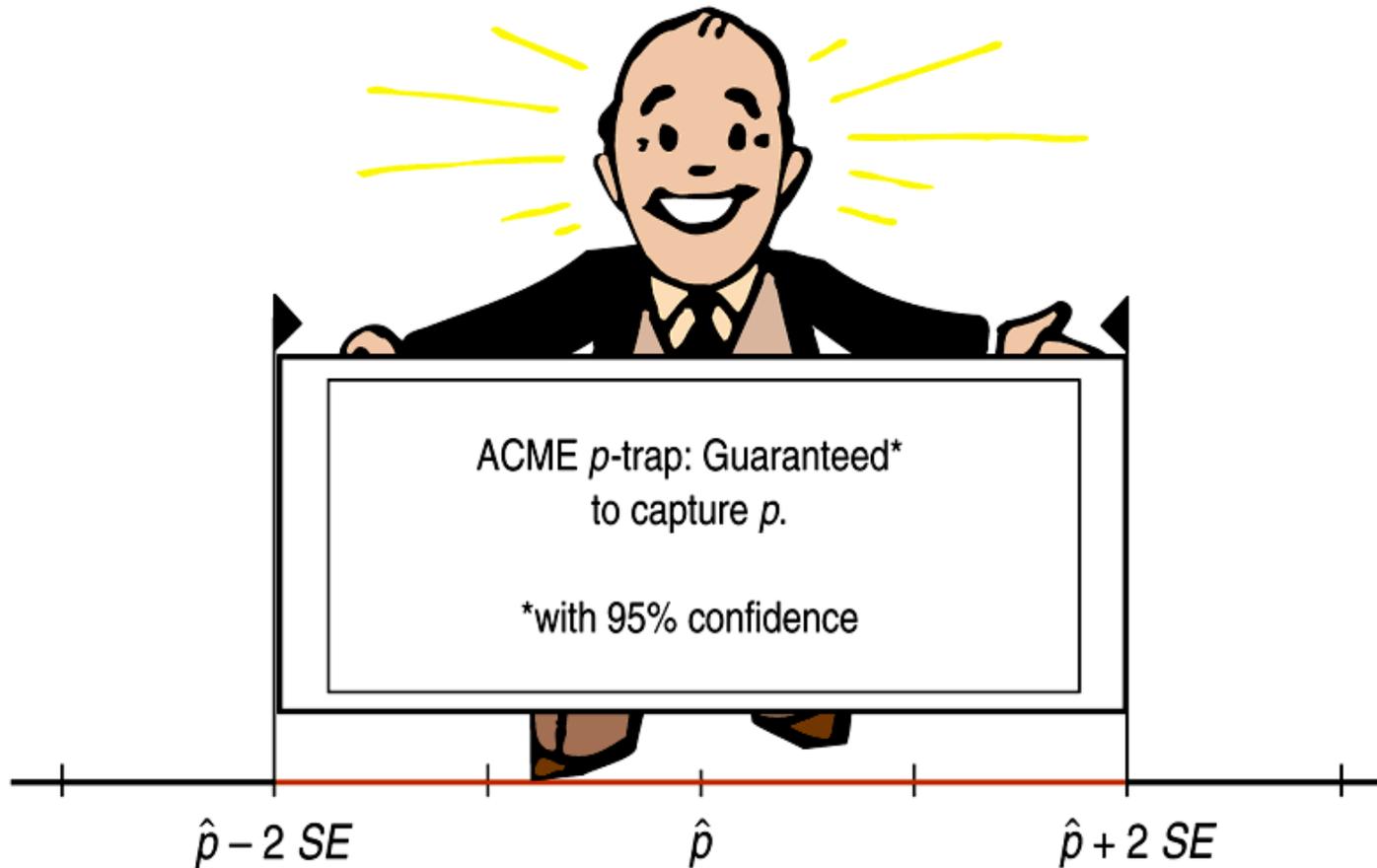
# Confidence Interval and 68-95-99.7% Rule

- By the 68-95-99.7% Rule, we know
  - about 68% of all samples will have  $\hat{p}$ 's within 1 *SE* of  $p$
  - about 95% of all samples will have  $\hat{p}$ 's within 2 *SEs* of  $p$
  - about 99.7% of all samples will have  $\hat{p}$ 's within 3 *SEs* of  $p$
- We can look at this from  $\hat{p}$ 's point of view...

# Confidence Interval and 68-95-99.7% Rule (cont.)

- Consider the 95% level:
  - There's a 95% chance that  $p$  is no more than 2  $SEs$  away from  $\hat{p}$ .
  - So, if we reach out 2  $SEs$ , we are 95% sure that  $p$  will be in that interval. In other words, if we reach out 2  $SEs$  in either direction of  $\hat{p}$ , we can be 95% confident that this interval contains the **true proportion (population proportion)**.
- This is called a 95% **confidence interval**.
- One side note: When we use a 95% confidence interval, we will be having a 5% significance level. We will learn more about it later in the course.

# Confidence Interval (cont.)

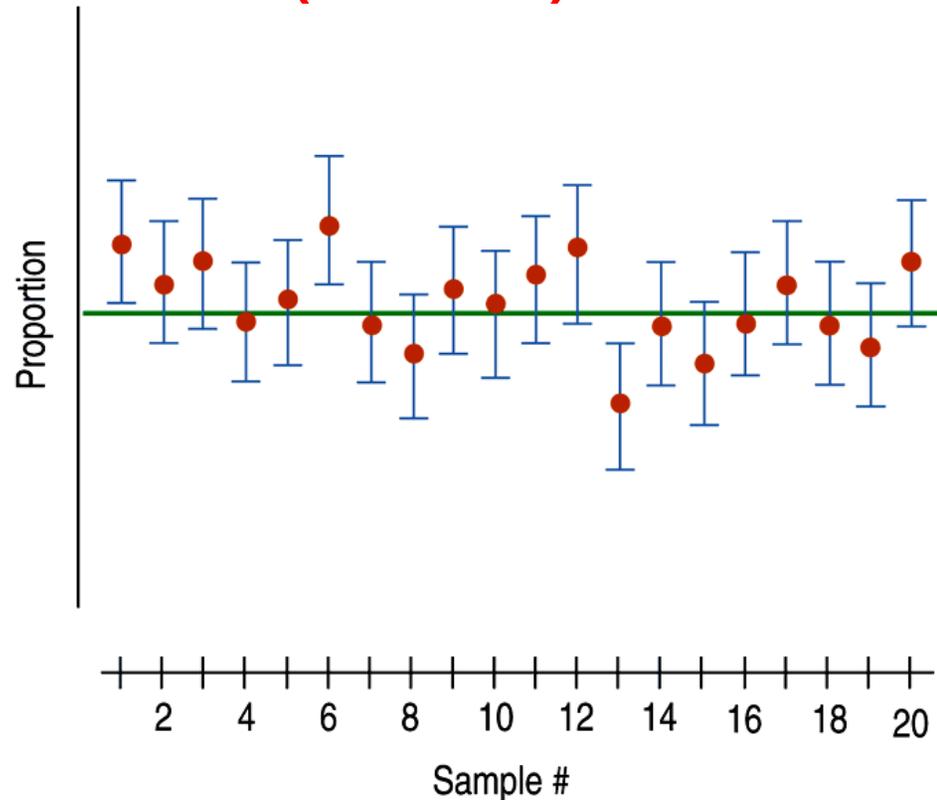


# What Does “95% Confidence” Really Mean?

- Each confidence interval uses a **sample statistic** to estimate a **population parameter**.
- But, since **samples vary**, the **statistics** we use, and thus the **confidence intervals** we construct, **vary** as well.

# What Does “95% Confidence” Really Mean? (cont.)

- The figure to the right shows that some of our confidence intervals capture the true proportion (the green horizontal line), while others do not:



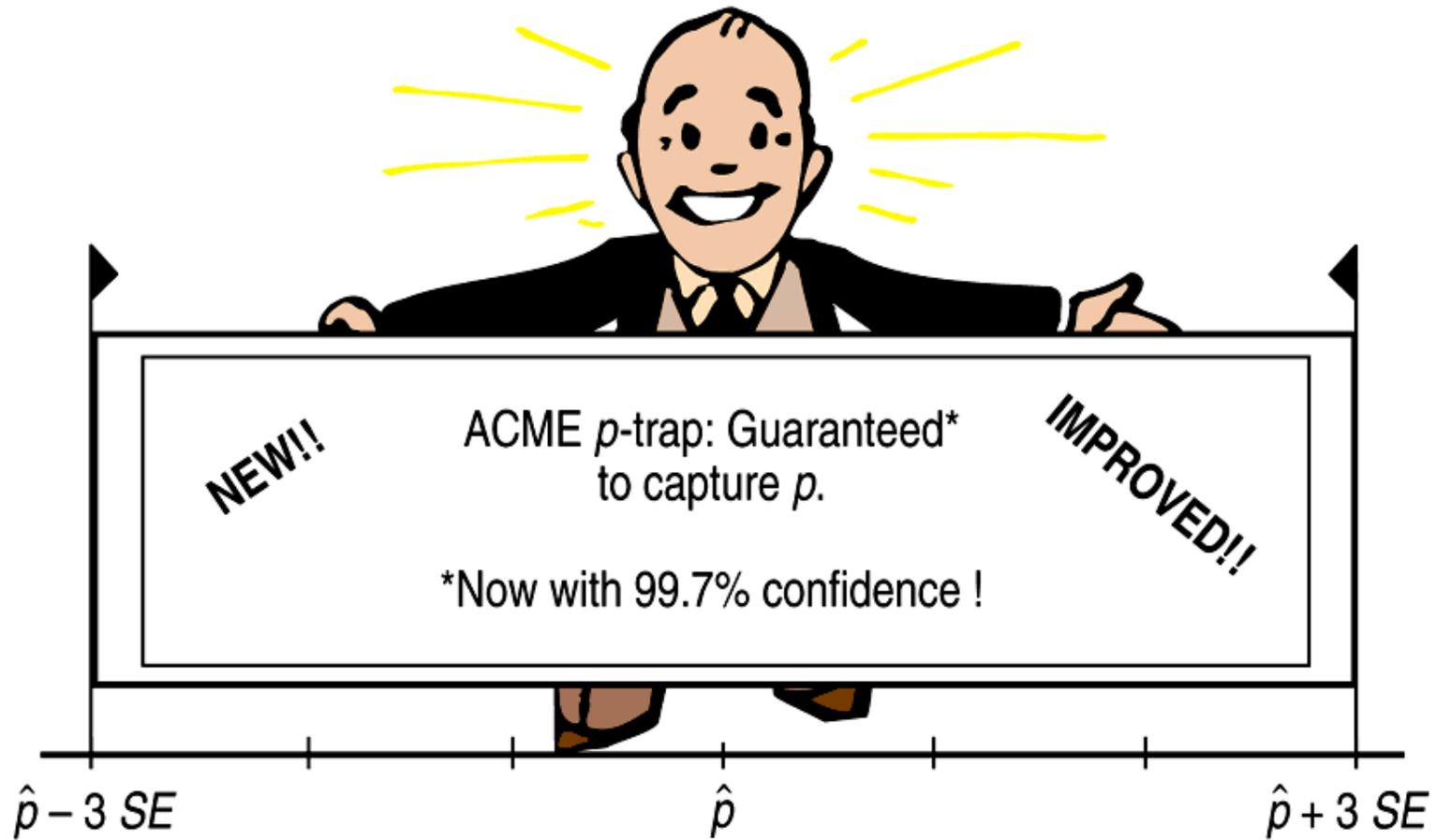
## What Does “95% Confidence” Really Mean? (cont.)

- Our confidence is in the *process of constructing the interval*, not in any one interval itself.
- Thus, we expect 95% of all 95% confidence intervals to contain the *true parameter* that they are estimating.

# Margin of Error: Certainty vs. Precision

- We can claim, with 95% confidence, that the interval  $\hat{p} \pm 2SE(\hat{p})$  contains the true population proportion.
  - The extent of the interval on either side of  $\hat{p}$  is called the margin of error ( $ME$ ).
- In general, confidence intervals have the form  $estimate \pm ME$ .
- The more confident we want to be, the larger our  $ME$  needs to be.

# Margin of Error: Certainty vs. Precision (cont.)



# Margin of Error: Certainty vs. Precision (cont.)

- To be **more confident**, we wind up **being less precise**.
  - We need more values in our confidence interval to be more certain.
- Because of this, every confidence interval is a balance between **certainty** and **precision**.
- The tension between certainty and precision is always there.
  - Fortunately, in most cases we can be both **sufficiently certain** and **sufficiently precise** to make useful statements.

# Margin of Error: Certainty vs. Precision (cont.)

- The choice of **confidence level** is somewhat arbitrary, but keep in mind this tension between **certainty** and **precision** when selecting your confidence level.
- The most commonly chosen **confidence levels** are 90%, 95%, and 99% (but any percentage can be used).
- Another side note: 95% confidence interval will give us a 5% significance level. Similarly, a 90% confidence interval will give us a 10% significance level. Thus, the most commonly chosen significance level would be 10%, 5%, and 1%.

# Critical Values

- The '2' in  $\hat{p} \pm 2SE(\hat{p})$  (our 95% confidence interval) came from the 68-95-99.7% Rule.
- Using a table or technology, we find that a more exact value for our 95% confidence interval is 1.96 instead of 2.

[Calculator hint: [2nd][Distr] 3:invNorm(0.025), where 0.025 (2.5%) is coming from dividing the significance level by two.]

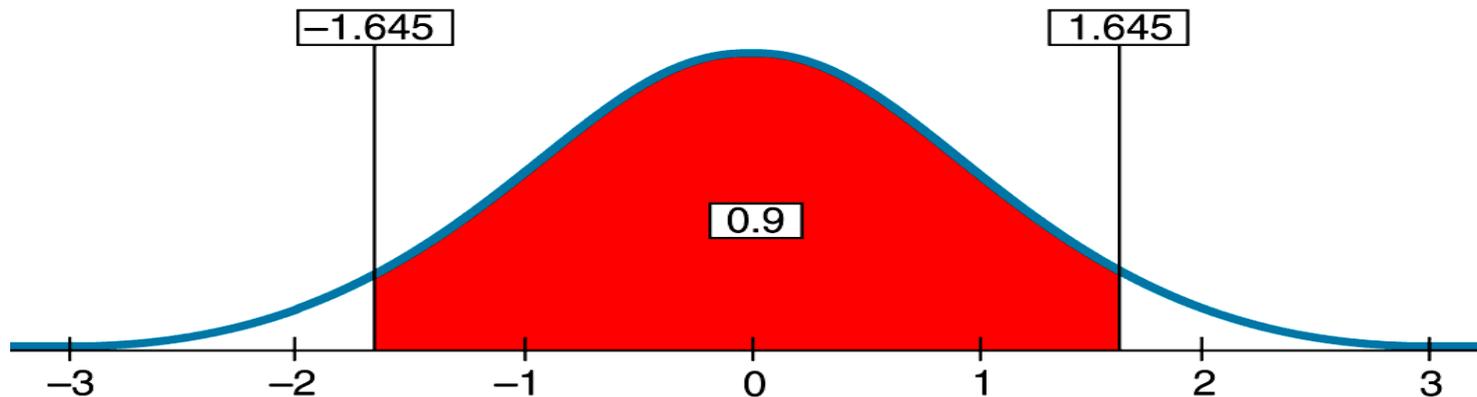
## Critical Values (cont.)

- We call 1.96 the **critical value** and denote it  $z^*$ .
- For any **confidence level**, we can find the corresponding **critical value**.
- The critical value can be obtained from the standard-normal table or technology, based on the chosen significance level.

Note: In this case, the significance level is 5% because the confidence level is 95%. To obtain the critical value, we divide the significance level by two for each tail.

## Critical Values (cont.)

Example: For a 90% confidence interval, the **critical value** is 1.645:



**Calculator hint:** You can obtain the critical value  $-1.645$  from your calculator TI-84+ by performing: `[2nd][Distr] 3:invNorm(0.05)`.

# Assumptions and Conditions

- All statistical models depend upon **assumptions**.
  - Different models depend upon different assumptions.
  - If those assumptions are not true, the model might be inappropriate and our conclusions based on it may be wrong.
- You can never be sure that an assumption is true, but you can often decide whether an assumption is plausible by checking a related **condition**.

## Assumptions and Conditions (cont.)

- Here are the assumptions and the corresponding conditions you must check before creating a confidence interval for a proportion:
- **Independence Assumption:** We first need to *Think* about whether the **Independence Assumption** is plausible. It's not one you can check by looking at the data. Instead, we check two conditions to decide whether independence is reasonable.

## Assumptions and Conditions (cont.)

- **Randomization Condition:** Were the data sampled at random or generated from a properly randomized experiment? Proper randomization can help ensure independence.
- **10% Condition:** Is the sample size no more than 10% of the population?
- **Sample Size Assumption:** The sample needs to be large enough for us to be able to use the CLT.
  - **Success/Failure Condition:** We must expect at least 5 successes and at least 5 failures.

# One-Proportion z-Interval

- When the conditions are met, we are ready to find the **confidence interval** for the **population proportion,  $p$** .
- The **confidence interval** is  $\hat{p} \pm z^* \times SE(\hat{p})$

where  $SE(\hat{p}) = \sqrt{\frac{\hat{p}\hat{q}}{n}}$

- The **critical value,  $z^*$** , depends on the particular confidence level,  $C$ , that you specify.

# How to Find A Confidence Interval for a Proportion?

## One-Proportion z-Interval Procedure

*Purpose* To find a confidence interval for a population proportion,  $p$

*Assumptions*

1. Simple random sample
2. The number of successes,  $x$ , and the number of failures,  $n - x$ , are both 5 or greater.

**STEP 1** For a confidence level of  $1 - \alpha$ , use Table II to find  $z_{\alpha/2}$ .

**STEP 2** The confidence interval for  $p$  is from

$$\hat{p} - z_{\alpha/2} \cdot \sqrt{\hat{p}(1 - \hat{p})/n} \quad \text{to} \quad \hat{p} + z_{\alpha/2} \cdot \sqrt{\hat{p}(1 - \hat{p})/n}$$

where  $z_{\alpha/2}$  is found in Step 1,  $n$  is the sample size, and  $\hat{p} = x/n$  is the sample proportion.

**STEP 3** Interpret the confidence interval.

# What Can Go Wrong?

## Don't Misstate What the Interval Means:

- Don't suggest that the parameter varies.
- Don't claim that other samples will agree with yours.
- Don't be certain about the parameter.
- Don't forget: It's the parameter (not the statistic).
- Don't claim to know too much.
- Do take responsibility (for the uncertainty).

# What Can Go Wrong? (cont.)

## Margin of Error Too Large to Be Useful:

- We can't be exact, but how precise do we need to be?
- One way to make the margin of error smaller is to reduce your level of confidence. (That may not be a useful solution.)
- You need to think about your margin of error when you design your study.
  - To get a narrower interval without giving up confidence, you need to have less variability.
  - You can do this with a larger sample...

## Violations of Assumptions:

- Watch out for biased sampling.
- Think about independence.

# What have we learned?

We have learned to:

1. Find and interpret a large-sample confidence interval for a population proportion.
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# Credit

Some of these slides have been adapted/modified in part/whole from the slides of the following textbooks.

- Weiss, Neil A., Introductory Statistics, 8th Edition
- Bock, David E., Stats: Data and Models, 3rd Edition