

STAT 216 Coursepack



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Montana State University

Melinda Yager
Jade Schmidt
Stacey Hancock

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Preface

This coursepack accompanies the textbook for STAT 216: Montana State Introductory Statistics with R, which can be found at <https://mtstateintrostats.github.io/IntroStatTextbook/>. The syllabus for the course (including the course calendar), data sets, and links to Canvas, Gradescope, and the MSU RStudio server can be found on the course webpage: <https://math.montana.edu/courses/s216/>. Other notes and review materials are linked in Canvas.

Each of the activities in this workbook is designed to target specific learning outcomes of the course, giving you practice with important statistical concepts in a group setting with instructor guidance. In addition to the in-class activities for the course, video notes are provided to aid in taking notes while you complete the required videos. Bring this workbook with you to class each class period, and take notes in the workbook as you would your own notes. A well-written completed workbook will provide an optimal study guide for exams!

All activities and labs in this coursepack will be completed during class time. Parts of each lab will be turned in on Gradescope. To aid in your understanding, read through the introduction for each activity before attending class each day.

STAT 216 is a 3-credit in-person course. In our experience, it takes six to nine hours per week outside of class to achieve a good grade in this class. By “good” we mean at least a C because a grade of D or below does not count toward fulfilling degree requirements. Many of you set your goals higher than just getting a C, and we fully support that. You need roughly nine hours per week to review past activities, read feedback on previous assignments, complete current assignments, and prepare for the next day’s class. A typical week in the life of a STAT 216 student looks like:

- *Prior to class meeting:*
 - Read assigned sections of the textbook, using the provided reading guides to take notes on the material.
 - Watch the provided videos, taking notes in the coursepack.
 - Read through the introduction to the day’s in-class activity.
 - Read through the week’s homework assignment and note any questions you may have on the content.
- *During class meeting:*
 - Work through the guided activity, in-class activity or weekly lab with your classmates and instructor, taking detailed notes on your answers to each question in the activity.
- *After class meeting:*
 - Complete any parts of the activity you did not complete in class.
 - Review the activity solutions in the Math and Stat Center, and take notes on key points.
 - Complete any remaining assigned readings for the week.
 - Complete the week’s homework assignment.

Basics of Data and Sampling Methods

1.1 Vocabulary Review and Key Topics

At the beginning of each module is a summary of key topics and new vocabulary terms for that module. As you read through the material in the textbook and watch the videos prior to class, look for these terms. Reference the definitions to guide your understanding.

1.1.1 Key topics

Module 1 introduces the foundations of data: observational units, types of variables, and how to collect sample data from a population of interest in a way that allows us to generalize our results back to the population.

1.1.2 Vocabulary

- **Data:** observations used to answer research questions.
- **Observational units (cases):** the subjects or entities on which data are collected.
 - The rows in a data set represent the observational units.
- **Sample size:** the number of observational units in a data set, denoted by n .
- **Variable:** the characteristics collected on each observational unit.
- **Types of variables:**
 - **Categorical:** cases are grouped into categories.
 - **Quantitative:** numerical measurements, where performing arithmetic operations makes sense.
- **Target population:** group of observational units of interest.
- **Sample:** subset of the population.
- **Statistic:** numerical value calculated on a sample; for categorical data, a proportion is calculated, for quantitative data, the mean is calculated.
- **Parameter:** numerical value of the entire population we are interested in; this is generally an unknown value.
- **Sampling methods:**
 - **Unbiased sampling method (e.g., a random sample):** on average, the sample will be representative of the target population; all observational units in the target population have the same chance of being selected.
 - **Biased sampling method (e.g., convenience sample):** on average, the sample will not be representative of the target population; some part of the target population will be over- or under-represented.
- **Type of sampling bias:**
 - **Selection bias:** method of sampling is biased; some part of the target population is over- or under-represented.

- **Non-response bias:** part of a pre-selected sample does not respond or cannot be reached.
- **Response bias:** responses are not truthful (poor/leading question phrasing, social desirability).

- **Generalization:** to what group of observational units can the results be applied to?
 - If an unbiased method of selection was used and there is no non-response or response bias, we can generalize the results to the target population.
 - If a biased method of selection was used or if non-response or response bias is present, we can only generalize the result to the sample or similar observational units.

1.2 Activity 1: Intro to Data Analysis and Sampling Bias

1.2.1 Learning outcomes

- Identify observational units, variables, and variable types in a statistical study.
- Creating a data set
- Identify biased sampling methods.

1.2.2 Terminology review

Statistics is the study of how best to collect, analyze, and draw conclusions from data. In the next few days of class, we will learn the building blocks of the semester. This week in class you will be introduced to the following terms:

- Observational units or cases
- Variables: categorical or quantitative
- Selection bias
- Response bias
- Non-response bias

For more on these concepts, read Chapter 1 and 2 in the textbook and review the Module 1 Vocabulary Review and Key Topics.

Notes on Observational Units and Variables

Further analysis of class data set

1. What are the observational units or cases for the data collected in class on day 1?
2. How many observations are reported in the data set? This is the **sample size**.
3. The header for each column in the data set describes each variable measured on the observational unit. For each column of data, fill in the following table identifying the type of each variable.
 - If the variable is categorical, indicate in the third column of the following table whether the variable is binary.
 - If the variable is quantitative, indicate in the forth column the units of measure used.

Column	Type of Variable	Binary?	Units?
Major			
Residency			
Num Credits			
Dominant hand			
Hand Span			
Grip strength dominant hand			
Grip strength non-dominant hand			

4. Review the completed data set with your class. Remember that when creating a data set for use in R it is important to use single words or an underscore between words. Each outcome must be written the same way each time to have consistency between responses. Do not include units of measure in the data set when reporting numerical values. Write down some issues found with the created class data set.

Notes on Sampling Methods and Types of bias

Types of bias

Complete Q5 together as a class:

5. A television station is interested in predicting whether or not local voters will pass a referendum to legalize marijuana for adult. The TV station asks its viewers to phone in and indicate whether they are in favor or opposed to the referendum. Of the 2241 viewers who phoned in, forty-five percent were opposed to legalizing marijuana.

Sample size:

Observational units sampled:

Target population:

Justify why there is selection bias in this study.

6. To determine if the proportion of out-of-state undergraduate students at Montana State University has increased in the last 10 years, a statistics instructor sent an email survey to 500 randomly selected current undergraduate students. One of the questions on the survey asked whether they had in-state or out-of-state residency. She only received 378 responses.

Sample size:

Observational units sampled:

Target population:

Justify why there is non-response bias in this study.

7. To gauge the interest of Bozeman City Voters in a new swimming pool, a local organization stood outside of the Bogart Pool in Bozeman, MT, during open hours. One of the questions they asked was, “Since the Bogart Pool is in such bad repair, don’t you agree that the city should fund a new pool?”

Sample size:

Observational units sampled:

Target population:

Justify why there is response bias in this study.

Justify why there is selection bias in this study.

1.2.3 Take-home messages

1. There are two types of variables: categorical (groups) and quantitative (numerical measures).
2. We will learn more about summarizing variable later in the semester. Categorical variables are summarized by calculating a proportion from the data and quantitative variables are summarized by finding the mean and the standard deviation.
3. There are three types of bias to be aware of when designing a sampling method: selection bias, non-response bias, and response bias.

1.2.4 Additional notes

Use this space to summarize your thoughts and take additional notes on today’s activity and material covered, and to write down the names and contact information of your teammates.

1.3 Activity 2: American Indian Address

1.3.1 Learning outcomes

- Explain why a sampling method is unbiased or biased.
- Identify biased sampling methods.
- Explain the purpose of random selection and its effect on generalization.

1.3.2 Terminology review

In this activity, we will examine unbiased and biased methods of sampling. Some terms covered in this activity are:

- Random sample
- Unbiased vs biased methods of selection
- Generalization

To review these concepts, see Chapter 2 in the textbook.

1.3.3 Class Preparation

Prior to the next class, complete questions 1–3.

1.3.4 American Indian Address

For this activity, you will read a speech given by Jim Becenti, a member of the Navajo American Indian tribe, who spoke about the employment problems his people faced at an Office of Indian Affairs meeting in Phoenix, Arizona, on January 30, 1947 (Moquin and Van Doren 1973). His speech is below:

It is hard for us to go outside the reservation where we meet strangers. I have been off the reservation ever since I was sixteen. Today I am sorry I quit the Santa Fe [Railroad]. I worked for them in 1912–13. You are enjoying life, liberty, and happiness on the soil the American Indian had, so it is your responsibility to give us a hand, brother. Take us out of distress. I have never been to vocational school. I have very little education. I look at the white man who is a skilled laborer. When I was a young man I worked for a man in Gallup as a carpenter's helper. He treated me as his own brother. I used his tools. Then he took his tools and gave me a list of tools I should buy and I started carpentering just from what I had seen. We have no alphabetical language.

We see things with our eyes and can always remember it. I urge that we help my people to progress in skilled labor as well as common labor. The hope of my people is to change our ways and means in certain directions, so they can help you someday as taxpayers. If not, as you are going now, you will be burdened the rest of your life. The hope of my people is that you will continue to help so that we will be all over the United States and have a hand with you, and give us a brotherly hand so we will be happy as you are. Our reservation is awful small. We did not know the capacity of the range until the white man come and say "you raise too much sheep, got to go somewhere else," resulting in reduction to a skeleton where the Indians can't make a living on it. For eighty years we have been confused by the general public, and what is the condition of the Navajo today? Starvation! We are starving for education. Education is the main thing and the only thing that is going to make us able to compete with you great men here talking to us.

By eye selection

1. Circle ten words in Jim Becenti's speech which are a representative sample of the length of words in the entire text. Describe your method for selecting this sample.
2. Fill in the table below with your selected words from the previous question and the length of each word (number of letters/digits in the word):

Observation	Word	Length
1		
2		
3		
4		
5		
6		
7		
8		
9		
10		

3. Calculate the mean (average) word length in your selected sample. Is this value a parameter or a statistic?

Notes on sampling

1.3.5 Class Activity

1. Report your mean word length from question 3 to your instructor. Your instructor will create a visualization of the distribution of results generated by your class. Draw a picture of the plot here. Include a descriptive x -axis label. Report the mean and standard deviation of the sample mean word lengths.

The plot created in question 1 is a sampling distribution of statistics. This sampling distribution plots the mean word length from many samples taken from the population of words.

2. The true mean word length of the population of all 359 words in the speech is 3.95 letters. Is this value a parameter or a statistic?

Where does the value of 3.95 fall in the plot given? Near the center of the distribution? In the tails of the distribution?

3. Based on the class discussion, would you say the sampling method used (“by-eye” selection) by the class is biased or unbiased? Justify your answer.
4. If the sampling method is biased, what type of sampling bias (selection, response, non-response) is present? What is the direction of the bias, i.e., does the method tend to overestimate or underestimate the population mean word length?

Random selection

Suppose instead of attempting to select a representative sample by eye (which did not work), each student used a random number generator to select a simple random sample of 10 words. A **simple random sample** relies on a random mechanism to choose a sample, without replacement, from the population, such that every sample of size 10 is equally likely to be chosen.

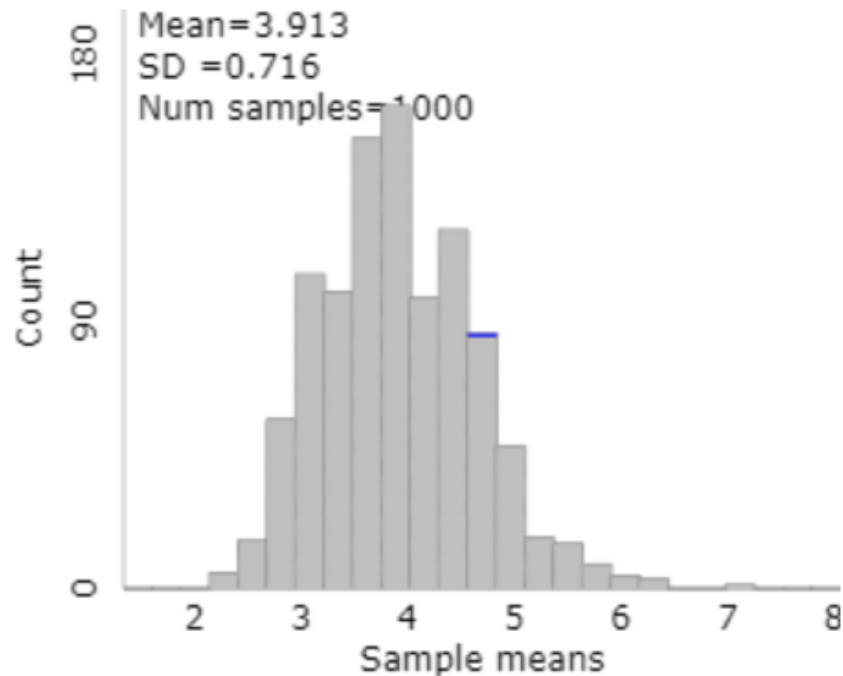
To use a random number generator to select a simple random sample, you first need a numbered list of all the words in the population, called a **sampling frame**. You can then generate 10 random numbers from the numbers 1 to 359 (the number of words in the population), and the chosen random numbers correspond to the chosen words in your sample.

5. Use the random number generator at <https://istats.shinyapps.io/RandomNumbers/> to select a simple random sample from the population of all 359 words in the speech.
 - Set “Choose Minimum” to 1 and “Choose Maximum” to 359 to represent the 359 words in the population (the sampling frame).
 - Set “How many numbers do you want to generate?” to 10 and ensure the “No” option is selected under “Sample with Replacement?”
 - Click “Generate”.

Fill in the table on the next page with the random numbers selected and use the Becenti.csv data file found on Canvas to determine each number’s corresponding word and word length (number of letters/digits in the word):

The following plot illustrates a sampling distribution of 1000 samples of size 10 selected at random from the sample.

Statistic: ☒ Mean ☐ Median ☐ *t*-statistic



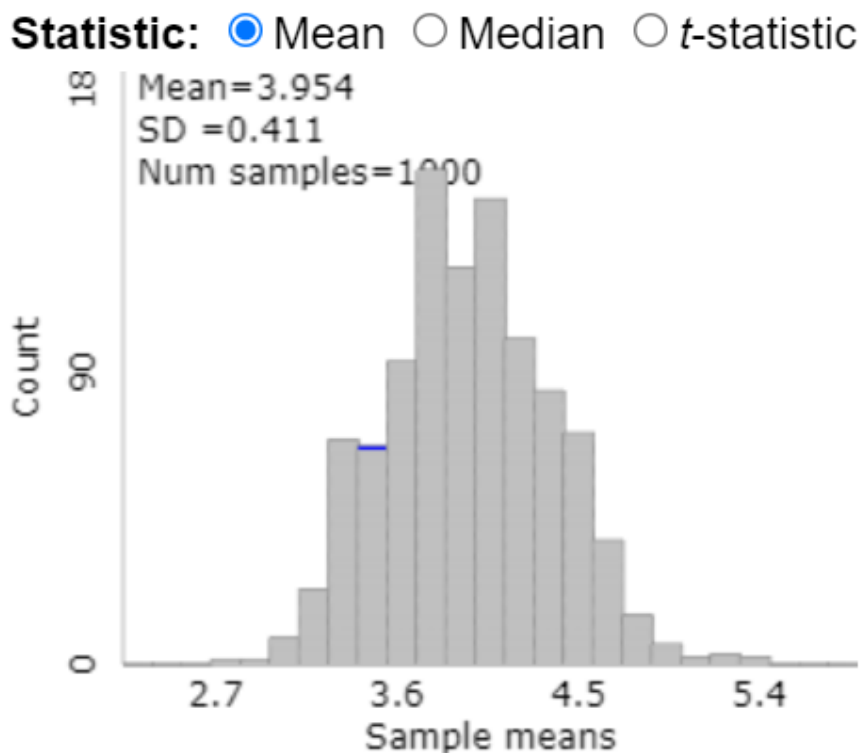
9. What is the center value (mean) of the distribution displayed above?
10. Explain why the sampling method of using a random number generator to generate a sample is a “better” method than choosing 10 words “by eye”.
11. Is random selection an unbiased method of selection? Explain your answer. Be sure to reference the plot from before Q9.

Effect of sample size

We will now consider the impact of sample size.

12. First, consider if each student had selected 30 words, instead of 10, by eye. Do you think this would make the plot from the previous activity centered on 3.95 (the true mean word length)? Explain your answer.

Now we will select 30 words instead of 10 words at random. The following plot illustrates a sampling distribution of 1000 samples of size 30 selected at random from the sample.



13. Compare the values of the standard deviation of the plots before question 9 and before question 13. Which plot shows the smallest standard deviation?
14. Using the evidence from your simulations, answer the following research questions:
- Does changing the sample size impact whether the sample estimates are unbiased? Explain your answer.
- Does changing the sample size impact the variability (spread) of sample estimates? Explain your answer.
15. What is the purpose of random selection of a sample from the population?

1.3.6 Take-home messages

1. When we use a biased method of selection, we will over or underestimate the parameter.
2. If the sampling method is biased, inferences made about the population based on a sample estimate will not be valid.
3. Random selection is an unbiased method of selection.
4. To determine if a sampling method is biased or unbiased, we compare the distribution of the estimates to the true value. We want our estimate to be on target or unbiased. When using unbiased methods of selection, the mean of the distribution matches or is very similar to our true parameter.
5. Random selection eliminates selection bias. However, random selection will not eliminate response or non-response bias.
6. The larger the sample size, the more similar (less variable) the statistics will be from different samples.
7. Sample size has no impact on whether a *sampling method* is biased or not. Taking a larger sample using a biased method will still result in a sample that is not representative of the population.

1.3.7 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

Probability

2.1 Vocabulary Review and Key Topics

2.1.1 Key topics

Module 2 introduces the concept of probability as a long-run relative frequency and demonstrates how to use hypothetical two-way tables to set up a probability problem and solve for unconditional and conditional probabilities.

2.1.2 Vocabulary

- **Probability** (of an event): the long-run proportion of times the event would occur if the random process were repeated indefinitely (under identical conditions).
- **Conditional probability** (of an event *given* another event): probability of an event calculated dependent on another event having occurred.
- **Probability notation:**
 - $P(A)$: the probability of event A .
 - * This is the probability of a single event, *unconditional* probability calculated out of the overall population.
 - $P(A^C)$: the probability of the **complement** of event A , or “ A complement”.
 - * This is the probability of the opposite of event A , or “not A ”.
 - * $P(A^C) = 1 - P(A)$
 - $P(A \text{ and } B)$: the probability of event A and B .
 - * This is the probability of an “and” event, *unconditional* probability calculated out of the overall population.
 - $P(A|B)$: the probability of event A given (conditional on) event B .
 - * This is a *conditional* probability calculated out of the total population for which event B occurred.

2.2 Activity 3: Probability Studies

2.2.1 Learning outcomes

- Recognize and simulate probabilities as long-run frequencies.
- Construct two-way tables to evaluate conditional probabilities.

2.2.2 Terminology review

In today's activity, we will cover two-way tables and probability. Some terms covered in this activity are:

- Proportions
- Probability
- Conditional probability
- Two-way tables

To review these concepts, see Chapter 23 in the textbook.

Notes on probability

The probability of an event is the long-run proportion of times the event would occur if the random process were repeated indefinitely (under identical conditions).

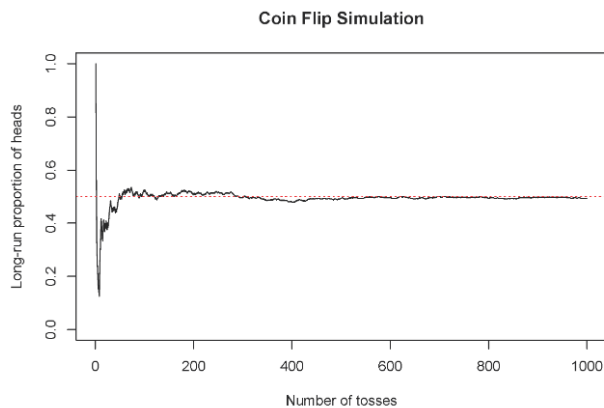
To calculate the probability of an event happening:

$$\text{probability} = \frac{\text{number of ways an event can happen}}{\text{total number of possible outcomes}}$$

For example, to calculate the probability of a coin flip landing on heads; there are only two outcomes (heads or tails) and only one possibility way to land on heads.

$$P(\text{heads}) = \frac{1}{2} = 0.5$$

The figure below shows the long-run proportion of times a simulated coin flip lands on heads on the y-axis, and the number of tosses on the x-axis. Notice how the long-run proportion starts converging to 0.5 as the number of tosses increases.



In today's activity we will discuss the probability of a single event, the probability of an "and" event, and the probability of a conditional event.

Probability notation

We will use the notation $P(\text{event})$ to represent the probability of an event and use letters to represent events. The following are notations for different probabilities where we are discussing event A and event B:

- $P(A)$ represents the probability of event A
- $P(A^C)$ represents the probability of the complement of event A
 - $P(A^C) = 1 - P(A)$
- $P(A \text{ and } B)$ represents the probability of events A and B
- $P(A|B)$ represents the probability of event A, given event B
- $P(B|A)$ represents the probability of event B, given event A

Probability questions

For the beginning of this activity we will start with discussing the probabilities associated with drawing a card from a standard card deck. In a card deck there are:

- 52 cards
- Half are red, half are black
- Four suits: spades, hearts, diamonds, and clubs
- Each suit has 13 cards: cards 2–10, ace, jack, queen, and king
- Let A represent the event that a card is an ace
- Let B represent the event that a card is red

To find the probability of selecting an ace, first start with determining how many aces are possible (four) and how many cards will we select from (total of 52).

Find the probability of selecting a card that is not an ace. This is the complement of event A.

Find the probability of selecting a red ace.

Find the probability of selecting an ace given that the card is red.

If a card drawn is an ace, what is the probability the card drawn is red.

Calculating probabilities from a two-way table

1. In 2014, the website FiveThirtyEight examined the works of Bob Ross to see what trends could be found. They determined that of all the paintings he created, 95% of them contained at least one “happy tree.” Of those works with a happy tree, 43% contained at least one “almighty mountain.” Of the paintings that did not have at least one happy tree, only 10% contained at least one almighty mountain.

Let A = Bob Ross painting contains a happy tree, and B = Bob Ross painting contains an almighty mountain

	A	A^C	Total
B	40850	500	41350
B^C	54150	4500	58650
Total	95000	5000	100000

- a. What is the probability that a randomly selected Bob Ross painting contains both a “happy tree” and an “almighty mountain”? Use appropriate probability notation.
- b. What is the probability that a selected Bob Ross painting without an “almighty mountain” contains a “happy tree.” Use appropriate probability notation.
- c. What is the probability that a selected Bob Ross painting does not contain a “happy tree” given it does not contain an “almighty mountain”. Use appropriate probability notation.

2. A recent study of population decline of white-tailed deer in Wyoming due to chronic wasting disease (Edmunds 2016) (CWD) reported that 35.4% of white-tailed deer have CWD. The survival rate of deer with CWD is 39.6% and the survival rate of deer without CWD is 80.1%.

Let A = the event a deer has CWD, and B = the event the deer survives.

- a. Identify what each numerical value given in the problem represents in probability notation.

$$0.354 =$$

$$0.396 =$$

$$0.801 =$$

- b. Create a hypothetical two-way table to represent the situation.

	A	A^C	Total
B			
B^C			
Total			100,000

- c. Find $P(A \text{ and } B)$. What does this probability represent in the context of the problem?
- d. Find the probability that a deer that has CWD does not survive. What is the notation used for this probability?
- e. What is the probability that a deer does not survive given they do not have CWD? What is the notation used for this probability?

3. Since the early 1980s, the rapid antigen detection test (RADT) of group A *streptococci* has been used to detect strep throat. A recent study of the accuracy of this test shows that the **sensitivity**, the probability of a positive RADT given the person has strep throat, is 86% in children, while the **specificity**, the probability of a negative RADT given the person does not have strep throat, is 92% in children. The **prevalence**, the probability of having group A strep, is 37% in children. (Stewart et al. 2014)

Let A = the event the child has strep throat, and B = the event the child has a positive RADT.

- a. Identify what each numerical value given in the problem represents in probability notation.

$$0.86 =$$

$$0.92 =$$

$$0.37 =$$

- b. Create a hypothetical two-way table to represent the situation.

	A	A^C	Total
B			
B^C			
Total			100,000

- c. Find $P(B)$. What does this probability represent in the context of the problem?
- d. Find the probability that a child with a positive RADT actually has strep throat. What is the notation used for this probability?
- e. What is the probability that a child does not have strep given that they have a positive RADT? What is the notation used for this probability?

2.2.3 Take home messages

1. Conditional probabilities are calculated dependent on a second variable. In probability notation, the variable following $|$ is the variable on which we are conditioning. The denominator used to calculate the probability will be the total for the variable on which we are conditioning.
2. When creating a two-way table we typically want to put the explanatory variable on the columns of the table and the response variable on the rows.
3. To fill in the two-way table, always start with the unconditional variable in the total row or column and then use the conditional probabilities to fill in the interior cells.

2.2.4 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

Exploring Categorical Data: Exploratory Data Analysis and Inference using Simulation-based Methods

3.1 Vocabulary Review and Key Topics

Review the Golden Ticket posted in the resources at the end of the coursepack for a summary of a single categorical variable.

3.1.1 Key topics

Module 3 introduces the steps of the statistical investigation process. We will conduct **exploratory data analysis** (summary statistics and plots) and simulation-based **inference** (hypothesis testing and confidence intervals) in the single categorical variable (one proportion) scenario.

- Notation for a sample proportion: \hat{p}
- Notation for a population proportion: π
- Types of plots for a single categorical variable:
 - Frequency bar plot
 - Relative frequency bar plot

Exploratory data analysis is step 3 of the statistical investigation process. We will then use simulation-based methods **to find evidence of an effect by finding a p-value** and **estimating how large the effect is by creating a confidence interval** in the one proportion (one categorical variable) scenario. These are steps 4 and 5 from the steps of the statistical investigation process.

Steps of the statistical investigation process

As we move through the semester we will work through the six steps of the statistical investigation process.

1. Ask a research question.
2. Design a study and collect data.
3. Summarize and visualize the data.
4. Use statistical analysis methods to draw inferences from the data.
5. Communicate the results and answer the research question.
6. Revisit and look forward.

3.1.2 Vocabulary

- **Summary measure:** a numerical quantity that summarizes data. Summary measures covered in STAT 216 include: single proportion, difference in proportions, single mean, paired mean difference, difference in means, correlation, and slope of a regression line.
 - For a single categorical variable, a proportion is calculated.

- **Summary statistic (point estimate):** the value of a numerical summary measure computed from *sample* data.
 - To interpret in context include:
 - * Summary measure (in context)
 - * Value of the statistic
- **Parameter of interest:** a numerical summary measure of the entire *population* in which we are interested.
 - The value of the parameter of interest is unknown (unless we have access to the entire population).
 - To write in context:
 - * Population word (true, long-run, population)
 - * Summary measure (depends on the type of data)
 - * Context
 - Observational units
 - Variable(s)
- For a single categorical variable, the category that we are counting the proportion of is generically called a “**success**”, with categories not a success labeled “**failure**”. Thus, a sample proportion is the “proportion of successes” in the sample: the total number of successes divided by the sample size (n).

Plotting one categorical variable

- **Frequency bar plot:** plots the count (frequency) of observational units in each level of a categorical variable. R code to create a frequency bar plot:

```
object %>% # Data set piped into...
ggplot(aes(x = variable)) + # This specifies the variable
geom_bar(stat = "count") + # Tell it to make a bar plot
labs(title = "Don't forget to title your plot!",
      # Give your plot a title
      x = "x-axis label", # Label the x axis
      y = "Frequency") # Label the y axis
```

- **Relative frequency bar plot:** plots the proportion (relative frequency) of observational units in each level of a categorical variable. R code to create a relative frequency bar plot:

```
object %>% # Data set piped into...
ggplot(aes(x = variable)) + # This specifies the variable
geom_bar(aes(y = after_stat(prop), group = 1)) + # Tell it to make a bar plot with proportions
labs(title = "Don't forget to title your plot!",
      # Give your plot a title
      x = "x-axis label", # Label the x axis
      y = "Relative Frequency") # Label the y axis
```

Inference

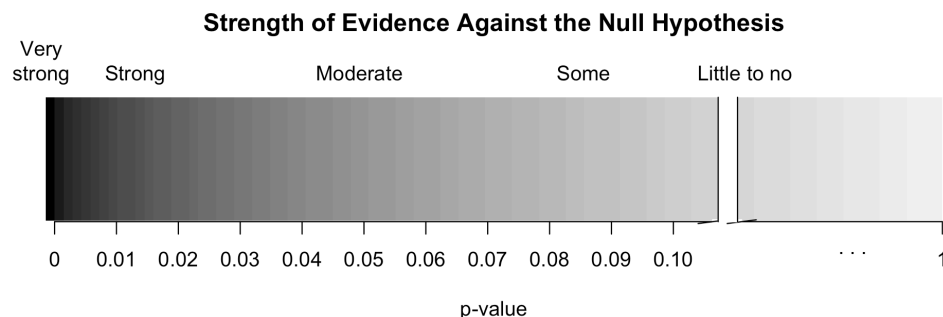
- **Sampling distribution** (of a statistic): the distribution of possible values of a statistic across repeated samples of the same size and under the same conditions.
 - We can create a *simulated* sampling distribution using simulation-based methods to simulate many samples, or we can mathematically model the sampling distribution (theory-based methods).

- **Hypothesis testing:** a formal statistical technique for evaluating two competing possibilities about a population: the null hypothesis and alternative hypothesis.
 - When we observe an effect in a sample, we would like to determine if this observed effect represents an actual effect in the population, or whether it was simply due to random chance.
 - A hypothesis test helps us answer the following question about the population: How strong is the *evidence* of an effect?
- **Null hypothesis:** typically represents a statement of “no difference”, “no effect”, or the status quo.
 - The null hypothesis is what we assume is true when calculating the p-value. Thus, we can never have evidence *for* the null hypothesis—we cannot “accept” a null hypothesis—we can only find evidence *against* the null hypothesis if the observed data is very unlikely to have occurred under the assumption that the null hypothesis is true.
- **Alternative hypothesis:** represents an alternative claim under consideration and is often represented by a range of possible values for the parameter of interest.
 - The alternative hypothesis is determined by the research question.
- **Hypotheses in notation for a single proportion:** In the hypotheses below, π_0 is the **null value**.

$$H_0 : \pi = \pi_0$$

$$H_A : \pi \left\{ \begin{array}{l} < \\ \neq \\ > \end{array} \right\} \pi_0$$

- **P-value:** the probability of the value of the observed sample statistic or a value more extreme, if the null hypothesis were true.
 - To write in context include:
 - * Statement about probability or proportion of samples
 - * Statistic (summary measure and value)
 - * Direction of the alternative
 - * Null hypothesis (in context)
- **Strength of evidence:** the p-value indicates the amount of evidence there is against the null hypothesis. The smaller the p-value the more evidence there is against the null hypothesis.



- **Conclusion** (to a hypothesis test): answers the research question. How much evidence is there in support of the alternative hypothesis?
 - To write in context include:
 - * Amount of evidence
 - * Parameter of interest
 - * Direction of the alternative hypothesis
- **Confidence interval**: an interval estimate for the parameter of interest; an interval of *plausible values* for the parameter.
 - A confidence interval helps us answer the following question about the population: How *large* is the effect?
 - To write in context include:
 - * How confident you are (e.g., 90%, 95%, 98%, 99%)
 - * Parameter of interest
 - * Calculated interval

Simulation-based inference for a single proportion

- **Conditions necessary to use simulation-based methods for inference for a single categorical variable**:
 - **Independence**: observational units must be independent of one another; the outcome of one observational unit should have no influence on the outcome of another.
- **Null distribution**: a sampling distribution of simulated sample statistics created under the assumption that the null hypothesis is true
- **Simulation-based methods to create the null distribution**: a process of using a computer program (e.g., R) to simulate many samples that we would expect based on the null hypothesis.

R code to use simulation methods for one categorical variable to find the p-value, `one_proportion_test` (from the `catstats` package), is shown below.

```
one_proportion_test(probability_success = xx, # Null hypothesis value
  sample_size = xx, # Enter sample size
  number_repetitions = 10000, # Enter number of simulations
  as_extreme_as = xx, # Observed statistic
  direction = "xx", # Specify direction of alternative hypothesis
  summary_measure = "proportion") # Reporting proportion or number of successes?
```

- **Bootstrapping**: creating a simulated sample of the same size as the original sample by sampling with replacement from the original sample.
- **Simulation-based methods to create the bootstrap distribution**: a process of using a computer program to simulate many bootstrapped samples.

R code to use simulation methods for one categorical variable to find a confidence interval, `one_proportion_bootstrap_CI` (from the `catstats` package), is shown below.

```
one_proportion_bootstrap_CI(sample_size = xx, # Sample size
  number_successes = xx, # Observed number of successes
  number_repetitions = 10000, # Number of bootstrap samples to use
  confidence_level = 0.95) # Confidence level as a decimal
```

- **Percentile method:** process to find the confidence interval from the bootstrap distribution.
 - A 90% confidence interval will be found between the 5th and 95th percentiles of the bootstrap distribution.
 - A 95% confidence interval will be found between the 2.5th and 97.5th percentiles of the bootstrap distribution.
 - A 99% confidence interval will be found between the 0.5th and 99.5th percentiles of the bootstrap distribution.

3.2 Activity 4: Helper-Hinderer Part 1 — Simulation-based Hypothesis Test

3.2.1 Learning outcomes

- Identify the two possible explanations (one assuming the null hypothesis and one assuming the alternative hypothesis) for a relationship seen in sample data.
- Given a research question involving a single categorical variable, construct the null and alternative hypotheses in words and using appropriate statistical symbols.
- Describe and perform a simulation-based hypothesis test for a single proportion.

3.2.2 Terminology review

In today's activity, we will work through a simulation-based hypothesis testing for a single categorical variable. Some terms covered in this activity are:

- Parameter of interest
- Null hypothesis
- Alternative hypothesis
- Simulation

To review these concepts, see Chapters 9 & 14 in your textbook.

3.2.3 Steps of the statistical investigation process

We will work through a five-step process to complete a hypothesis test for a single proportion, first introduced in the activity in week 1.

- **Ask a research question** that can be addressed by collecting data. What are the researchers trying to show?
- **Design a study and collect data.** This step involves selecting the people or objects to be studied and how to gather relevant data on them.
- **Summarize and visualize the data.** Calculate summary statistics and create graphical plots that best represent the research question.
- **Use statistical analysis methods to draw inferences from the data.** Choose a statistical inference method appropriate for the data and identify the p-value and/or confidence interval after checking assumptions. In this study, we will focus on using randomization to generate a simulated p-value.
- **Communicate the results and answer the research question.** Using the p-value and confidence interval from the analysis, determine whether the data provide statistical evidence against the null hypothesis. Write a conclusion that addresses the research question.

Notes on Hypothesis Testing

3.2.4 Helper-Hinderer

A study by Hamblin, Wynn, and Bloom reported in Nature (Hamblin, Wynn, and Bloom 2007) was intended to check young kids' feelings about helpful and non-helpful behavior. Non-verbal infants ages 6 to 10 months were shown short videos with different shapes either helping or hindering the climber. As a class we will watch this short video to see how the experiment was run: <https://youtu.be/anCaGBsBOxM>. Researchers were hoping to assess: Are non-verbal infants more likely to choose the helper toy? In the study, of the 16 infants age 6 to 10 months, 14 chose the *helper* toy and 2 chose the *hinderer* toy.

- Observational units:
- Variable:
- Type of variable:
- Success:

Ask a research question

- Research question:

Design a study and collect data

Before using statistical inference methods, we must check that the cases are independent. The sample observations are independent if the outcome of one observation does not influence the outcome of another. One way this condition is met is if data come from a simple random sample of the target population.

1. Are the cases independent? Justify your answer.

R code

For almost all activities and labs it will be necessary to upload the provided R script file from Canvas for that day. Your instructor will highlight a few steps in uploading files to and using RStudio.

The following are the steps to upload the necessary R script file for this activity:

- Download the Activity R script file from Canvas.
- Click “Upload” in the “Files” tab in the bottom right window of RStudio. In the pop-up window, click “Choose File”, and navigate to the folder where the Activity R script file is saved (most likely in your downloads folder). Click “Open”; then click “Ok”.
- You should see the uploaded file appear in the list of files in the bottom right window. Click on the file name to open the file in the Editor window (upper left window).

Notice that the first three lines of code contain a prompt called `library`. Packages needed to run functions in R are stored in directories called libraries. When using the MSU RStudio server, all the packages needed for the class are already installed. We simply must tell R which packages we need for each R script file. We use the prompt `library` to load each **package** (or library) needed for each activity. Note, these `library` lines MUST be run each time you open a R script file in order for the functions in R to work.

- Highlight and run lines 1–3 to load the packages needed for this activity. Notice the use of the `#` symbol in the R script file. This symbol is not part of the R code. It is used by these authors to add comments to the R code and explain what each call is telling the program to do.

R will ignore everything after a `#` symbol when executing the code. Refer to the instructions following the `#` symbol to understand what you need to enter in the code.

```
library(tidyverse)
library(ggplot2)
library(catstats)
```

Throughout activities, we will often include the R code you would use in order to produce output or plots. These “code chunks” appear in gray. In the code chunk below, we demonstrate how to read the data set into R using the `read.csv()` function. The line of code shown below (line 7 in the R script file) reads in the data set and names the data set `infants`.

Summarize and visualize the data

The following code reads in the data set and gives the number of infants in each level of the variable, whether the infant chose the helper or the hinderer.

- Highlight and run lines 7 and 8 to check that you get the same counts as shown below

```
# Read in data set
infants <- read.csv("https://math.montana.edu/courses/s216/data/infantchoice.csv")
infants %>% count(choice) # Count number in each choice category
```

```
#>      choice  n
#> 1  helper 14
#> 2 hinderer  2
```

The following formula is used to calculate the proportion of successes in the sample.

$$\hat{p} = \frac{\text{number of successes}}{\text{total number of observational units}}$$

2. Using the R output and the formula given, calculate the summary statistic (sample proportion) to represent the research question. Recall that **choosing the helper toy** is a considered a success. Use appropriate notation.

To visually display this data we can use either a frequency bar plot or a relative frequency bar plot.

- Enter the variable name **choice** for **variable** in the R code to create the frequency bar plot.
- Note the name of the title is given in line 16 and includes the **type of plot**, **observational units**, and **variable name**.
- Highlight and run lines 13–19 to create the plot

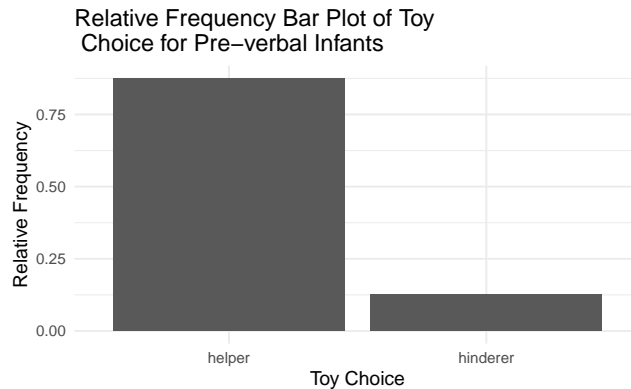
```
infants %>% # Data set piped into...
  ggplot(aes(x = variable)) + # This specifies the variable
  geom_bar(stat = "count") + # Tell it to make a bar plot
  labs(title = "Frequency Bar Plot of Toy Choice for Pre-verbal Infants",
        # Give your plot a title
        x = "Toy Choice", # Label the x axis
        y = "Frequency") # Label the y axis
```

3. Sketch the frequency bar plot created below.

We could also choose to display the data as a proportion in a **relative frequency** bar plot. To find the relative frequency, the count in each level of **choice** is divided by the sample size. This calculation is the sample proportion for each level of **choice**. Notice that in the following code we told R to create a bar plot with proportions.

- In the R script file, highlight and run lines 23–29 to create the relative frequency bar plot.

```
infants %>% # Data set piped into...
  ggplot(aes(x = choice)) + # This specifies the variable
  geom_bar(aes(y = after_stat(prop), group = 1)) + # Tell it to make a bar plot with proportions
  labs(title = "Relative Frequency Bar Plot of Toy \n Choice for Pre-verbal Infants",
        # Give your plot a title
        x = "Toy Choice", # Label the x axis
        y = "Relative Frequency") # Label the y axis
```



4. Which features in the relative frequency bar plot are the same as the frequency bar plot? Which are different?

We cannot assess whether infants are more likely to choose the helper toy based on the statistic and plot alone. The next step is to analyze the data by using a hypothesis test to discover if there is evidence against the null hypothesis.

Use statistical analysis methods to draw inferences from the data

When performing a hypothesis test, we must first identify the null hypothesis. The null hypothesis is written about the parameter of interest, or the value that summarizes the variable in the population.

The parameter of interest is a statement about what we want to find about the population. The following must be included when writing the parameter of interest.

- Population word (true, long-run, population)
- Summary measure (depends on the type of data)
- Context
 - Observational units
 - Variable(s)

Parameter of interest:

If the children are just randomly choosing the toy, we would expect half (0.5) of the infants to choose the helper toy. This is the null value for our study.

Null Hypothesis (in words):

The notation used for a population proportion (or probability, or true proportion) is π . Since this summarizes a population, it is a parameter. When writing the **null hypothesis** in notation, we set the parameter equal to the null value, $H_0 : \pi = \pi_0$.

Null Hypothesis (in notation):

The **alternative hypothesis** is the claim to be tested and the direction of the claim (less than, greater than, or not equal to) is based on the research question.

- Based on the research question from question 1, are we testing that the parameter is greater than 0.5, less than 0.5 or different than 0.5?

Alternative hypothesis (in words):

Alternative hypothesis (in notation):

Remember that when utilizing a hypothesis test, we are evaluating two competing possibilities. For this study the **two possibilities** are either...

- The true proportion of infants who choose the helper is 0.5 and our results just occurred by random chance; or,
- The true proportion of infants who choose the helper is greater than 0.5 and our results reflect this.

Notice that these two competing possibilities represent the null and alternative hypotheses.

We will now simulate one sample of a **null distribution** of sample proportions. The null distribution is created under the assumption the null hypothesis is true. In this case, we assume the true proportion of infants who choose the helper is 0.5, so we will create 10000 (or more) different simulations of 16 infants under this assumption.

Let's think about how to use a coin to create one simulation of 16 infants under the assumption the null hypothesis is true. Let heads equal infant chose the helper toy and tails equal infant chose the hinderer toy.

5. How many times would you flip a coin to simulate the sample of infants?
6. Flip a coin 16 times recording the number of times the coin lands on heads. This represents one simulated sample of 16 infants randomly choosing the toy. Calculate the proportion of coin flips that resulted in heads.
7. Is the value from question 6 closer to 0.5, the null value, or closer to the sample proportion, 0.875?

Report the number of coin flips you got as indicated by your instructor.

8. Sketch the graph created by your instructor of each student's proportion of heads out of 16 coin flips.

9. Circle the observed statistic (value from question 2) on the distribution shown above. Where does this statistic fall in this distribution: Is it near the center of the distribution (near 0.5) or in one of the tails of the distribution?
10. Is the observed statistic likely to happen or unlikely to happen if the true proportion of infants who choose the helper is 0.5? Explain your answer using the plot.

In the next class, we will continue to assess the strength of evidence against the null hypothesis by using a computer to simulate 10000 samples when we assume the null hypothesis is true.

3.2.5 Take-home messages

1. Two types of plots are used for plotting categorical variables: frequency bar plots, relative frequency bar plots.
2. In a hypothesis test we have two competing hypotheses, the null hypothesis and the alternative hypothesis. The null hypothesis represents either a skeptical perspective or a perspective of no difference or no effect. The alternative hypothesis represents a new perspective such as the possibility that there has been a change or that there is a treatment effect in an experiment.
3. In a simulation-based test, we create a distribution of possible simulated statistics for our sample if the null hypothesis is true. Then we see if the calculated observed statistic from the data is likely or unlikely to occur when compared to the null distribution.
4. To create one simulated sample on the null distribution for a sample proportion, spin a spinner with probability equal to π_0 (the null value), n times or draw with replacement n times from a deck of cards created to reflect π_0 as the probability of success. Calculate and plot the proportion of successes from the simulated sample.

3.2.6 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

3.3 Activity 5: Helper-Hinderer (continued)

3.3.1 Learning outcomes

- Describe and perform a simulation-based hypothesis test for a single proportion.
- Interpret and evaluate a p-value for a simulation-based hypothesis test for a single proportion.
- Explore what a p-value represents

3.3.2 Steps of the statistical investigation process

In today's activity we will continue with steps 4 and 5 in the statistical investigation process. We will continue to assess the Helper-Hinderer study from last class.

- **Ask a research question** that can be addressed by collecting data. What are the researchers trying to show?
- **Design a study and collect data.** This step involves selecting the people or objects to be studied and how to gather relevant data on them.
- **Summarize and visualize the data.** Calculate summary statistics and create graphical plots that best represent the research question.
- **Use statistical analysis methods to draw inferences from the data.** Choose a statistical inference method appropriate for the data and identify the p-value and/or confidence interval after checking assumptions. In this study, we will focus on using randomization to generate a simulated p-value.
- **Communicate the results and answer the research question.** Using the p-value and confidence interval from the analysis, determine whether the data provide statistical evidence against the null hypothesis. Write a conclusion that addresses the research question.

3.3.3 Helper-Hinderer

In class today, we will revisit the study on infants as described below.

A study by Hamblin, Wynn, and Bloom reported in *Nature* (Hamblin, Wynn, and Bloom 2007) was intended to check young kids' feelings about helpful and non-helpful behavior. Non-verbal infants ages 6 to 10 months were shown short videos with different shapes either helping or hindering the climber. As a class we will watch this short video to see how the experiment was run: <https://youtu.be/anCaGBsBOxM>. Researchers were hoping to assess: Are non-verbal infants more likely to choose the helper toy? In the study, of the 16 infants age 6 to 10 months, 14 chose the *helper* toy and 2 chose the *hinderer* toy.

1. Report the sample proportion (summary statistic) calculated in the previous activity.
2. Write the alternative hypothesis in words in context of the problem. Remember the direction we are testing is dependent on the research question.

Today, we will use the computer to simulate a null distribution of 10000 different samples of 16 infants, plotting the proportion who chose the helper in each sample, based on the assumption that the true proportion of infants who choose the helper is 0.5 (or that the null hypothesis is true).

To use the computer simulation, we will need to enter the

- assumed “probability of success” (π_0),
 - “sample size” (the number of observational units or cases in the sample),
 - “number of repetitions” (the number of samples to be generated - typically we use 10000),
 - “as extreme as” (the observed statistic), and
 - the “direction” (matches the direction of the alternative hypothesis).
3. What values should be entered for each of the following into the one proportion test to create 10000 simulations?
- Probability of success (null value):
 - Sample size (n):
 - Number of repetitions (typically use 10000 simulations):
 - As extreme as (value of statistic):
 - Direction ("greater", "less", or "two-sided"):

We will use the `one_proportion_test()` function in R (in the `catstats` package) to simulate the null distribution of sample proportions and compute a p-value. Using the provided R script file, fill in the values/words for each `xx` with your answers from question 3 in the one proportion test to create a null distribution with 10000 simulations. Then highlight and run lines 1–16.

```
one_proportion_test(probability_success = xx, # Null hypothesis value
  sample_size = xx, # Enter sample size
  number_repetitions = 10000, # Enter number of simulations
  as_extreme_as = xx, # Observed statistic
  direction = "xx", # Specify direction of alternative hypothesis
  summary_measure = "proportion") # Reporting proportion or number of successes?
```

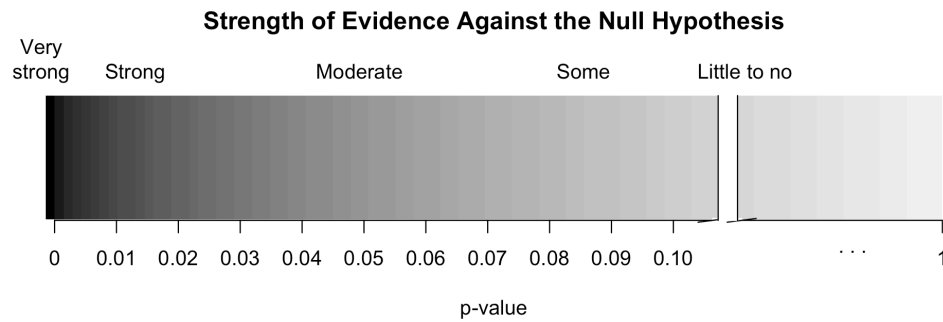
4. Sketch the null distribution created from the R code here.

Notes on the null distribution

5. Circle the observed statistic (value from question 1) on the distribution you drew in question 4. Where does this statistic fall in the null distribution: Is it near the center of the distribution (near 0.5) or in one of the tails of the distribution?
6. Is the observed statistic likely to happen or unlikely to happen if the true proportion of infants who choose the helper is 0.5? Explain your answer using the plot.
7. Using the simulation, what is the proportion of simulated samples that generated a sample proportion at the observed statistic or greater, if the true proportion of infants who choose the helper is 0.5? *Hint:* Look under the simulation.

Notes on the p-value

The value in question 7 is the **p-value**. The smaller the p-value, the more evidence we have against the null hypothesis.



Interpret the p-value

The p-value measures the probability that we observe a sample proportion as extreme as what was seen in the data or more extreme (matching the direction of the H_A) IF the null hypothesis is true. This is a conditional probability, calculated dependent on the null hypothesis being true. Represented in probability notation:

$$P(\text{statistic or more extreme} | \text{the null hypothesis is true})$$

p-value interpretation:

Communicate the results and answer the research question

When we write a conclusion we answer the research question by stating how much evidence there is in support of the alternative hypothesis.

Conclusion:

Generalization

8. To what group of observational units can the results be generalized to?

3.3.4 Take-home messages

1. The null distribution is created based on the assumption the null hypothesis is true. We compare the sample statistic to the distribution to find the likelihood of observing this statistic.
2. The p-value measures the probability of observing the sample statistic or more extreme (in direction of the alternative hypothesis) if the null hypothesis is true.
3. The smaller the p-value of the test, the more evidence there is **against** the null hypothesis.

3.3.5 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

3.4 Activity 6: Helper-Hinderer — Simulation-based Confidence Interval

3.4.1 Learning outcomes

- Use bootstrapping to find a confidence interval for a single proportion.
- Interpret a confidence interval for a single proportion.

3.4.2 Terminology review

In today's activity, we will introduce simulation-based confidence intervals for a single proportion. Some terms covered in this activity are:

- Parameter of interest
- Bootstrapping
- Confidence interval

To review these concepts, see Chapters 10 & 14 in your textbook.

3.4.3 Helper-Hinderer

In the last class, we found very strong evidence that the true proportion of infants who will choose the helper character is greater than 0.5. But what *is* the true proportion of infants who will choose the helper character? We will use this same study to estimate this parameter of interest by creating a confidence interval.

As a reminder: A study by Hamblin, Wynn, and Bloom reported in *Nature* (Hamblin, Wynn, and Bloom 2007) was intended to check young kids' feelings about helpful and non-helpful behavior. Non-verbal infants ages 6 to 10 months were shown short videos with different shapes either helping or hindering the climber. Researchers were hoping to assess: Are non-verbal infants more likely to choose the helper toy? In the study, of the 16 infants age 6 to 10 months, 14 chose the *helper* toy and 2 chose the *hinderer* toy.

A **point estimate** (our observed statistic) provides a single plausible value for a parameter. However, a point estimate is rarely perfect; usually there is some error in the estimate. In addition to supplying a point estimate of a parameter, a next logical step would be to provide a plausible *range* of values for the parameter. This plausible range of values for the population parameter is called an **interval estimate** or **confidence interval**.

Activity intro

1. What is the value of the point estimate (sample statistic)?
2. If we took another random sample of 16 infants, would we get the exact same point estimate? Explain why or why not.

In today's activity, we will use bootstrapping to find a 95% confidence interval for π , the parameter of interest.

Notes on Confidence Intervals

Use statistical analysis methods to draw inferences from the data

3. Write out the parameter of interest in words, in context of the study. What does π represent?

To create the null distribution we flipped a coin 16 times to simulate infants randomly choosing the helper toy with a probability of 50%.

4. Why can't we use a coin to simulate the bootstrap distribution.

To create the bootstrap distribution.

- First we would label the cards to represent the sample statistic: 14 helper and 2 hinderer.
 - Sample with replacement 16 times
5. Using the cards provided by your instructor, create one bootstrap sample. Report your simulated sample proportion on the whiteboard.

To use the computer simulation to create a bootstrap distribution, we will need to enter the

- “sample size” (the number of observational units or cases in the sample),
 - “number of successes” (the number of cases that choose the helper character),
 - “number of repetitions” (the number of samples to be generated), and
 - the “confidence level” (which level of confidence are we using to create the confidence interval).
6. What values should be entered for each of the following into the simulation to create the bootstrap distribution of sample proportions to find a 95% confidence interval?
 - Sample size (n):
 - Number of successes:
 - Number of repetitions (at least 10000):
 - Confidence level (as a decimal):

We will use the `one_proportion_bootstrap_CI()` function in R (in the `catstats` package) to simulate the bootstrap distribution of sample proportions and calculate a confidence interval. Using the provided R script file, fill in the values/words for each `xx` with your answers from question 6 in the one proportion bootstrap confidence interval (CI) code to create a bootstrap distribution with 10000 simulations. Then highlight and run lines 1–9.

```
one_proportion_bootstrap_CI(sample_size = xx, # Sample size
                             number_successes = xx, # Observed number of successes
                             number_repetitions = 10000, # Number of bootstrap samples to use
                             confidence_level = xx) # Confidence level as a decimal
```

7. Sketch the bootstrap distribution created below.

Notes on the bootstrap distribution

95% Confidence Interval:

Interpretation of the 95% confidence interval in context.

Communicate the results and answer the research question

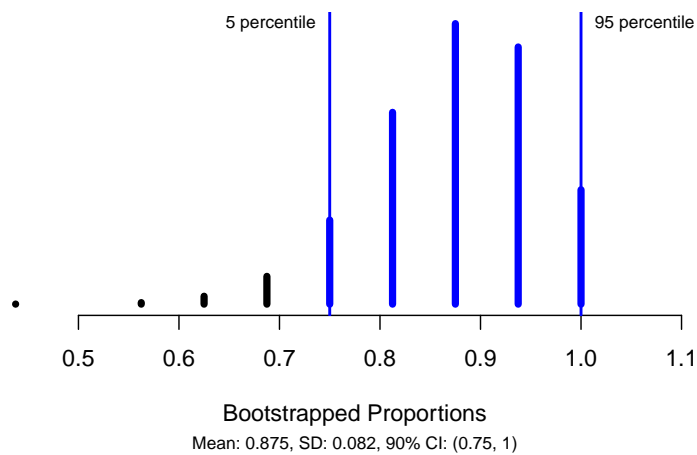
8. Is the value 0.5 (the null value) in the 95% confidence interval?

Explain how this indicates that the p-value provides strong evidence against the null.

Effect of confidence level

9. Suppose instead of finding a 95% confidence interval, we found a 90% confidence interval. Would you expect the 90% confidence interval to be narrower or wider? Explain your answer.
10. The following R code produced the bootstrap distribution with 10000 simulations that follows. Circle the value that changed in the code.

```
one_proportion_bootstrap_CI(sample_size = 16, # Sample size
                             number_successes = 14, # Observed number of successes
                             number_repetitions = 10000, # Number of bootstrap samples to use
                             confidence_level = 0.90) # Confidence level as a decimal
```



11. Report both the 95% confidence interval and the 90% confidence interval (question 10). Is the 90% confidence interval narrower or wider than the 95% confidence interval?

Concluding paragraph

In many of our studies we will write a paragraph summarizing the results of the study.

Researchers were interested if infants observe social cues and would be more likely to choose the helper toy. In a sample of 16 infants, 14 chose the helper toy. A simulation null distribution with 10000 simulations was created in RStudio. The p-value was found by calculating the proportion of simulations in the null distribution at the sample statistic of 0.875 and greater. This resulted in a p-value of 0.0024. We would observe a sample proportion of 0.875 or greater with a probability of 0.0024 IF we assume the true proportion of non-verbal infants who would choose the helper toy is 0.5. Based on this p-value, there is very strong evidence that the true proportion of infants age 6 to 10 months who will choose the helper toy is greater than 0.5. In addition, a 95% confidence interval was found for the parameter of interest. We are 95% confident that the true proportion of infants age 6 to 10 months who will choose the helper toy is between 0.75 and 1. The results of this study can be generalized to the sample of non-verbal infants as the researchers did not select a random sample.

3.4.4 Take-home messages

1. The goal in a hypothesis test is to assess the strength of evidence for an effect, while the goal in creating a confidence interval is to determine how large the effect is. A **confidence interval** is a range of *plausible* values for the parameter of interest.
2. A confidence interval is built around the point estimate or observed calculated statistic from the sample. This means that the sample statistic is always the center of the confidence interval. A confidence interval includes a measure of sample to sample variability represented by the **margin of error**.
3. In simulation-based methods (bootstrapping), a simulated distribution of possible sample statistics is created showing the possible sample-to-sample variability. Then we find the middle X percent of the distribution around the sample statistic using the percentile method to give the range of values for the confidence interval. This shows us that we are $X\%$ confident that the parameter is within this range, where X represents the level of confidence.
4. When the null value is within the confidence interval, it is a plausible value for the parameter of interest; thus, we would find a larger p-value for a hypothesis test of that null value. Conversely, if the null value is NOT within the confidence interval, we would find a small p-value for the hypothesis test and strong evidence against this null hypothesis.
5. To create one simulated sample on the bootstrap distribution for a sample proportion, label n cards with the original responses. Draw with replacement n times. Calculate and plot the resampled proportion of successes.

3.4.5 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

Inference for a Single Categorical Variable: Theory-based Methods

4.1 Vocabulary Review and Key Topics

Review the Golden Ticket posted in the resources at the end of the coursepack for a summary of a single categorical variable.

4.1.1 Key topics

Module 4 introduces theory-based inference methods (hypothesis testing and confidence intervals) for a single categorical variable. We also explore what “confidence level” means and which parts of a study impact the width of a confidence interval and the p-value.

- Theory-based methods should give the same results as simulation-based methods if the sample size is large enough. For a single categorical variable, the sample size is large enough if the success-failure condition is met.
- If repeated samples of the same size are taken from the population, 95% of samples will create a 95% confidence interval that contains the value of the parameter of interest.

4.1.2 Vocabulary

- **Theory-based methods:** when specific conditions are met, the distribution of sample statistics if we were to repeatedly sample from the population can be fit with a theoretical distribution.
- **Conditions for the sampling distribution of \hat{p} to follow an approximate normal distribution:**
 - **Independence:** the sample’s observations are independent, e.g., are from a simple random sample. (*Remember:* This also must be true to use simulation-based methods!)
 - **Large enough sample size:** Success-failure condition: we *expect* to see at least 10 successes and 10 failures in the sample, $n\pi \geq 10$ and $n(1 - \pi) \geq 10$. Since π is typically unknown, we consider this condition to be met if we observe at least 10 successes and 10 failures in our data set: $n\hat{p} \geq 10$ and $n(1 - \hat{p}) \geq 10$.
- **Standard normal distribution:** a theoretical distribution that is bell-shaped, centered on the mean of zero, and has a standard deviation of one, denoted in notation by $N(0, 1)$.
- **Standard error of a statistic:** an estimated standard deviation of the statistic as it would vary across repeated samples of the same size under the same conditions.
 - The standard error tells us about how far we would expect an observed sample statistic to fall from the true parameter value for which it is estimating, on average.
- **Standardized statistic:** calculation to standardize the sample statistic in order to compare the standardized value to the theoretical distribution.
 - Calculated by subtracting the null value from the sample statistic, then dividing by the standard error:

$$\frac{\text{statistic} - \text{null value}}{\text{standard error}}$$

- Measures the number of standard errors the sample statistic is above (if positive) or below (if negative) the null value.
- **Standard error of the sample proportion assuming the null is true:** measures the how far each possible sample proportion is from the true proportion, on average, and is calculated using the null value:

$$SE_0(\hat{p}) = \sqrt{\frac{\pi_0 \times (1 - \pi_0)}{n}}$$

- **Standardized sample proportion:** standardized statistic for a single categorical variable calculated using:

$$Z = \frac{\hat{p} - \pi_0}{SE_0(\hat{p})},$$

If the conditions for the sampling distribution of \hat{p} to follow an approximate normal distribution are met, and if the true value of π is equal to the null value of π_0 , the standardized sample proportion, Z , will have an approximate *standard normal* distribution.

- The theory-based **p-value** for hypothesis testing involving proportions can be found in R by using the **pnorm** function to find the probability of the observed standardized statistic or one more extreme (in the direction of H_A). This probability is the area under a *standard normal distribution* at or more extreme than the observed standardized statistic.
 - Enter the value of the standardized statistic for **xx**.
 - If a “greater than” alternative, change **lower.tail = TRUE** to **FALSE**.
 - If a two-sided test, multiply by 2.

```
pnorm(xx, lower.tail=TRUE)
```

- **Margin of error:** half the width of the confidence interval. For a single proportion, the margin of error is:

$$ME = z^* \times SE(\hat{p})$$

where z^* is the **multiplier**, corresponding to the desired confidence level found from the standard normal distribution. For example, for a 95% confidence level, the middle 95% of the standard normal distribution falls between $-z^* = -1.96$ and $z^* = 1.96$.

- **Standard error of the sample proportion for a confidence interval** (not assuming the null is true):

$$SE(\hat{p}) = \sqrt{\frac{\hat{p} \times (1 - \hat{p})}{n}}$$

- To find the endpoints of a confidence interval, add and subtract the margin of error to the sample statistic. The confidence interval for a population proportion is:

$$\hat{p} \pm ME$$

- R code to find the **multiplier** for the confidence interval using theory-based methods involving proportions.
 - **qnorm** will give you the multiplier using the standard normal distribution.
 - Enter the percentile for the given level of confidence (e.g., 0.975 for a 95% confidence level).

```
qnorm(percentile, lower.tail=TRUE)
```

4.2 Activity 7: Handedness of Male Boxers

4.2.1 Learning outcomes

- Describe and perform a theory-based hypothesis test for a single proportion.
- Check the appropriate conditions to use a theory-based hypothesis test.
- Calculate and interpret the standardized sample proportion.
- Interpret and evaluate a p-value for a theory-based hypothesis test for a single proportion.
- Use the normal distribution to find the p-value.

4.2.2 Terminology review

In this activity, we will introduce theory-based hypothesis tests for a single categorical variable. Some terms covered in this activity are:

- Parameter of interest
- Standardized statistic
- Normal distribution
- p-value

To review these concepts, see Chapter 11 & 14 in your textbook.

Activities from module 3 covered simulation-based methods for hypothesis tests involving a single categorical variable. This activity covers theory-based methods for testing a single categorical variable.

4.2.3 Handedness of male boxers

Left-handedness is a trait that is found in about 10% of the general population. Past studies have shown that left-handed men are over-represented among professional boxers (Richardson and Gilman 2019). Is there evidence that there is an over-prevalence of left-handed fighters? In this random sample of 500 professional male boxers, 81 were left-handed.

- Observational units:
- Variable:
- Type of variable:
- Success:

4.2.4 Summary statistics review

- Download the R file for today's activity from Canvas
- Upload the file to the R server
- Run lines 1–15 to load the needed packages and the data set and create a plot of the data

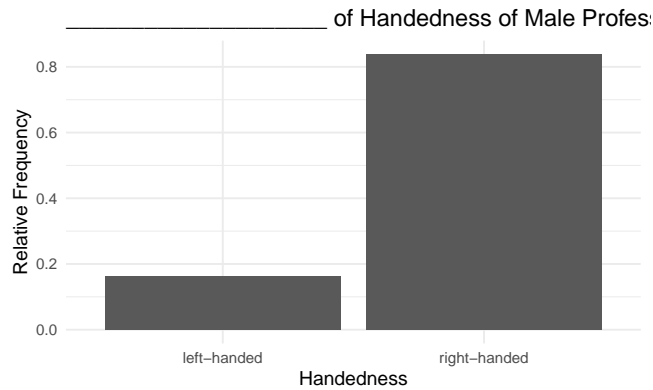
```
# Read in data set
boxers <- read.csv("https://math.montana.edu/courses/s216/data/Male_boxers_sample.csv")
boxers %>% count(Stance) # Count number in each Stance category
```

```
#>      Stance    n
#> 1 left-handed  81
#> 2 right-handed 419
```

```

boxers %>% # Data set piped into...
  ggplot(aes(x = Stance)) + # This specifies the variable
  geom_bar(aes(y = after_stat(prop), group = 1)) + # Tell it to make a bar plot with proportions
  labs(title = "_____ of Handedness of Male Professional Boxers",
        # Give your plot a title
        x = "Handedness", # Label the x axis
        y = "Relative Frequency") # Label the y axis

```



1. What type of plot was created of these data?

Hypotheses and summary statistics

2. Write out the parameter of interest in words, in context of the study.
3. Write out the null hypothesis **in words**.
4. Write out the alternative hypothesis **in notation**.
5. Calculate the value of the summary statistic (sample proportion) for this study. Use proper notation.

Theory-based methods

The sampling distribution of a single proportion — how that proportion varies from sample to sample — can be mathematically modeled using the normal distribution if certain conditions are met.

Conditions for the sampling distribution of \hat{p} to follow an approximate normal distribution:

- **Independence:** The sample's observations are independent, e.g., are from a simple random sample. (*Remember:* This also must be true to use simulation methods!)
- **Large enough sample size:** Success-failure condition: We *expect* to see at least 10 successes and 10 failures in the sample, $n\hat{p} \geq 10$ and $n(1 - \hat{p}) \geq 10$.

Additional notes on Theory-based methods

- Verify that the independence condition is satisfied.
- Verify that the sample size is large enough.

To calculate the standardized statistic we use the general formula

$$Z = \frac{\text{point estimate} - \text{null value}}{SE_0(\text{point estimate})}.$$

For a single categorical variable the standardized sample proportion is calculated using

$$Z = \frac{\hat{p} - \pi_0}{SE_0(\hat{p})},$$

where the standard error is calculated using the null value:

$$SE_0(\hat{p}) = \sqrt{\frac{\pi_0 \times (1 - \pi_0)}{n}}$$

For this study, the null standard error of the sample proportion is calculated using the null value, 0.1.

$$SE_0(\hat{p}) = \sqrt{\frac{0.1 \times (1 - 0.1)}{500}} = 0.013$$

Each sample proportion of male boxers that are left-handed is 0.013 from the true proportion of male boxers that are left-handed, on average.

Label the standard normal distribution shown below with the null value as the center value (below the value of zero). Label the tick marks to the right of the null value by adding 1 standard error to the null value to represent 1 standard error, 2 standard errors, and 3 standard errors from the null. Repeat this process to the left of the null value by subtracting 1 standard error for each tick mark.

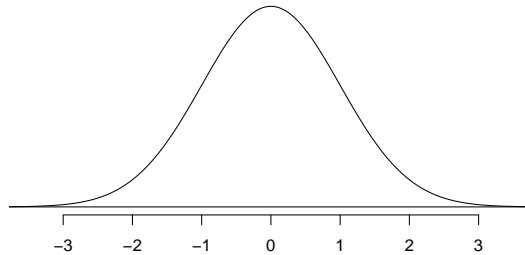


Figure 4.1: Standard Normal Curve

6. Using the null standard error of the sample proportion, calculate the standardized sample proportion (Z). Mark this value on the standard normal distribution above.

The standardized statistic is used as a ruler to measure how far the sample statistic is from the null value. Essentially, we are converting the sample proportion into a measure of standard errors to compare to the standard normal distribution.

The standardized statistic measures the *number of standard errors the sample statistic is from the null value*.

Interpretation of the standardized sample proportion:

We will use the `pnorm()` function in R to find the p-value. In the code below, notice that we used `lower.tail = FALSE` to find the p-value. R will calculate the p-value *greater* than the value of the standardized statistic.

Notes:

- Use `lower.tail = TRUE` when doing a left-sided test.
- Use `lower.tail = FALSE` when doing a right-sided test.
- To find a two-sided p-value, use a left-sided test for negative Z or a right-sided test for positive Z , then multiply the value found by 2 to get the p-value.

```
pnorm(4.769, # Enter value of standardized statistic
      m=0, s=1, # Using the standard normal mean = 0, sd = 1
      lower.tail=FALSE) # Gives a p-value greater than the standardized statistic
```

7. Report the p-value obtained from the R output.

8. Write a conclusion based on the p-value.

9. To what group of observational units can the results be generalized to?

Impacts on the P-value

Suppose that we want to show that the true proportion of male boxers **differs** from that in the general population.

10. Write out the alternative hypothesis in notation for this new research question.

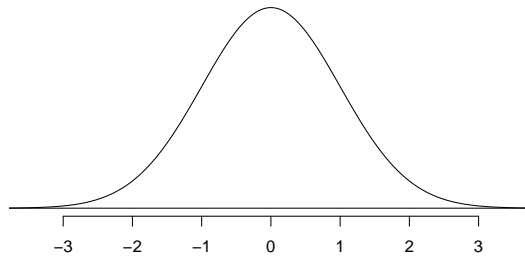


Figure 4.2: Standard Normal Curve

11. How would this impact the p-value? Would the p-value be larger or smaller?

Suppose instead of 500 male boxers the researchers only took a sample of 300 male boxers and found the same proportion ($\hat{p} = 0.162$) of male boxers that are left-handed. Since we are still assuming the same null value, 0.1, the standard error would be calculated as below:

$$SE_0(\hat{p}) = \sqrt{\frac{0.1(1 - 0.1)}{300}} = 0.017$$

.

The standardized statistic for this new sample is calculated below:

$$Z = \frac{0.162 - 0.1}{0.017} = 3.64$$

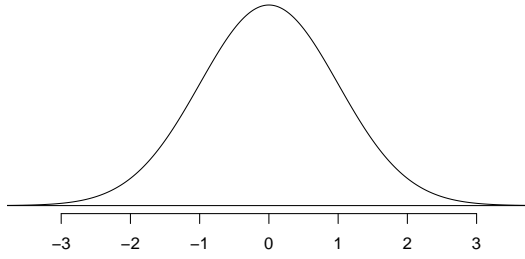


Figure 4.3: Standard Normal Curve

12. How does the decrease in sample size affect the p-value?

Suppose another sample of 500 male boxers was taken and 68 were found to be left-handed. Since we are still assuming the same null value, 0.1, the standard error would be calculated as before:

$$SE_0(\hat{p}) = \sqrt{\frac{0.1(1 - 0.1)}{500}} = 0.013$$

.

The standardized statistic for this new sample is calculated below:

$$Z = \frac{0.136 - 0.1}{0.013} = 2.769$$

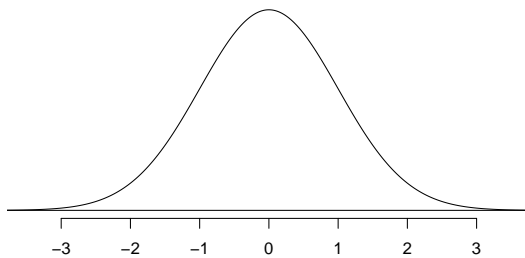


Figure 4.4: Standard Normal Curve

13. How does a statistic closer to the null value affect the p-value?

14. Summarize how each of the following affected the p-value:

- a) Switching to a two-sided test.
- b) Using a smaller sample size.
- c) Using a sample statistic closer to the null value.

4.2.5 Take-home messages

- 1. Both simulation and theory-based methods can be used to find a p-value for a hypothesis test. In order to use theory-based methods we need to check that both the independence and the success-failure conditions are met.
- 2. The standardized statistic measures how many standard errors the statistic is from the null value. The larger the standardized statistic the more evidence there is against the null hypothesis.
- 3. The p-value for a two-sided test is approximately two times the value for a one-sided test. A two-sided test provides less evidence against the null hypothesis.
- 4. The larger the sample size, the smaller the sample to sample variability. This will result in a larger standardized statistic and more evidence against the null hypothesis.
- 5. The farther the statistic is from the null value, the larger the standardized statistic. This will result in a smaller p-value and more evidence against the null hypothesis.

4.2.6 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

4.3 Activity 8: Confidence interval and what confidence means

4.3.1 Learning outcomes

- Explore what confidence means
- Interpret the confidence level
- Explore impact of sample size, direction of the alternative hypothesis, and value of the sample statistic on the p-value.

4.3.2 Terminology review

In this activity, we will explore what being 95% confidence means. Some terms covered in this activity are:

- Parameter of interest
- Two-sided vs. one-sided tests
- Confidence level

4.3.3 Handedness of male boxers continued

In today's activity, we will use the male boxer study to look at what confidence means.

Left-handedness is a trait that is found in about 10% of the general population. Past studies have shown that left-handed men are over-represented among professional boxers (Richardson and Gilman 2019). Is there evidence that there is an over-prevalence of left-handed fighters? In this random sample of 500 professional male boxers, 81 were left-handed.

```
# Read in data set
boxers <- read.csv("https://math.montana.edu/courses/s216/data/Male_boxers_sample.csv")
boxers %>% count(Stance) # Count number in each Stance category
```

```
#>      Stance    n
#> 1 left-handed  81
#> 2 right-handed 419
```

What does *confidence* mean?

In the interpretation of a 95% confidence interval, we say that we are 95% confident that the parameter is within the confidence interval. Why are we able to make that claim? What does it mean to say “we are 95% confident”?

1. In the last activity we found very strong evidence that the true proportion of male professional boxers that are left-handed is greater than 0.1. As a class, determine a plausible value for the true proportion of male boxers that are left-handed. *Note: we are making assumptions about the population here. This is not based on our calculated data, but we will use this applet to better understand what happens when we take many, many samples from this believed population.*
 2. Go to this website, <http://www.rossmanchance.com/ISIApplets.html> and choose ‘Simulating Confidence Intervals’. In the input on the left-hand side of the screen enter the value from question 1 for π (the true value), 500 for n , and 100 for ‘Number of intervals’. Click ‘sample’.
- In the graph on the bottom right, click on a green dot. Write down the confidence interval for this sample given on the graph on the left. Does this confidence interval contain the true value chosen in question 1?

- Now click on a red dot. Write down the confidence interval for this sample. Does this confidence interval contain the true value chosen in question 1?
 - How many intervals out of 100 contain π , the true value chosen in question 1? *Hint:* This is given to the left of the graph of green and red intervals.
3. Click on ‘sample’ nine more times. Write down the ‘Running Total’ for the proportion of intervals that contain π .
 4. Change the confidence level to 90%. What happened to the width of the intervals?
 5. Write down the **Running Total** for the proportion of intervals that contain π using a 90% confidence level.

Interpretation of the level of confidence:

Notes on theory-based confidence intervals

To calculate a theory-based 95% confidence interval for π , we will first find the **standard error** of \hat{p} by plugging in the value of \hat{p} for π in $SD(\hat{p})$:

$$SE(\hat{p}) = \sqrt{\frac{\hat{p} \times (1 - \hat{p})}{n}}$$

Note that we do not include a “0” subscript, since we are not assuming a null hypothesis.

Calculate the standard error of the sample proportion to find a 95% confidence interval.

We will calculate the margin of error and confidence interval later in this activity. **The margin of error (ME)** is the value of the z^* multiplier times the standard error of the statistic.

$$ME = z^* \times SE(\hat{p})$$

The z^* multiplier is the percentile of a standard normal distribution that corresponds to our confidence level. If our confidence level is 95%, we find the Z values that encompass the middle 95% of the standard normal distribution. If 95% of the standard normal distribution should be in the middle, that leaves 5% in the tails, or 2.5% in each tail.

The `qnorm()` function in R will tell us the z^* value for the desired percentile (in this case, $95\% + 2.5\% = 97.5\%$ percentile).

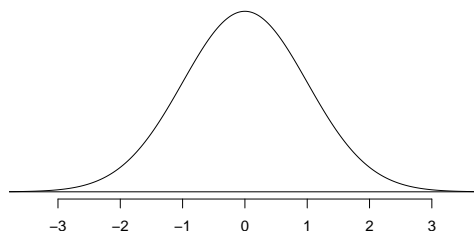


Figure 4.5: Standard Normal Curve

The following code will find the z^* value for a 95% confidence interval.

```
qnorm(c(0.025, 0.975), lower.tail = TRUE) # Multiplier for 95% confidence interval
```

Calculate the margin of error for the 95% confidence interval.

To find the confidence interval, we will add and subtract the **margin of error** to the point estimate:

point estimate \pm margin of error

$$\hat{p} \pm z^* \times SE(\hat{p})$$

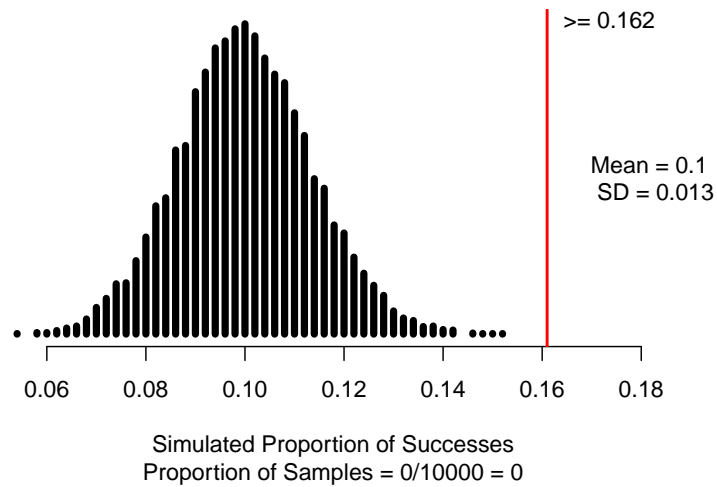
Calculate the 95% confidence interval for the parameter of interest.

6. Interpret the 95% confidence **interval** in the context of the problem.

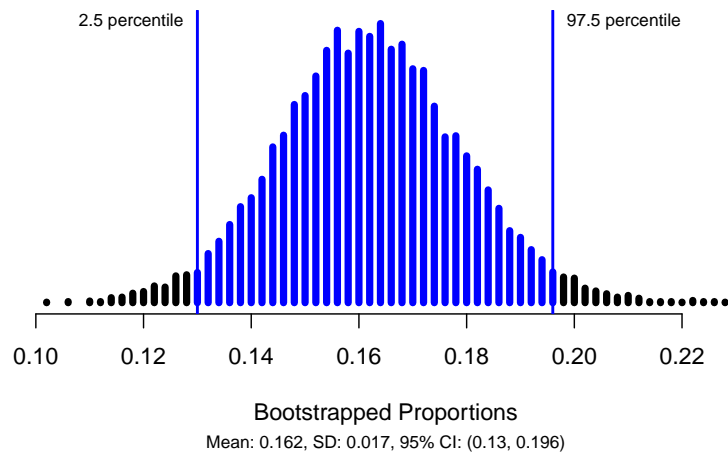
Simulation methods

We could also use simulation-based methods to analyze these data.

```
one_proportion_test(probability_success=0.1,  
  sample_size=500,  
  number_repetitions=10000,  
  as_extreme_as=0.162,  
  direction="greater",  
  summary_measure="proportion")
```



```
one_proportion_bootstrap_CI(sample_size = 500,
                             number_successes = 81,
                             number_repetitions = 10000,
                             confidence_level = 0.95)
```



7. Explain why the results for simulation methods and theory-based methods are similar.

Effect of sample size on the width of the confidence interval

How would an decrease in sample size impact the width of the confidence interval? Suppose instead of 500 male boxers the researchers only took a sample of 300 male boxers and found the same proportion ($\hat{p} = 0.162$) of male boxers that are left-handed.

The standard error of the sample proportion for this study with the smaller sample size is:

$$SE(\hat{p}) = \sqrt{\frac{0.162 \times (1 - 0.162)}{300}} = 0.0213$$

8. Is the standard error of the sample proportion for this study smaller or larger than the value calculated earlier?

Recall that the z^* multiplier is 1.96 for a 95% confidence interval.

9. Calculate the 95% confidence interval for this study with the smaller sample size.

The width of the confidence interval is found by calculating the difference between the upper value and the lower value.

$$\text{width of CI} = \text{upper CI value} - \text{lower CI value}$$

10. Compare the interval found in question 9 to the interval calculated prior to question 6.

- Did the center of the interval change?
- Calculate the width of the interval with the smaller sample size.
- Calculate the width of the interval from prior to question 6.
- Which interval is wider?

The margin of error represents half the width of the confidence interval since we add and subtract the margin of error to the value of the sample statistic.

$$\text{width of CI} = 2 \times \text{ME}$$

11. Using the width of the interval with the smaller sample size calculated in question 10, calculate the margin of error.

12. What impact does decreasing the sample size have on the width of the confidence interval?

4.3.4 Take-home messages

1. If repeat samples of the same size are selected from the population, approximately 95% of samples will create a 95% confidence interval that contains the parameter of interest.
2. The calculation of the confidence interval uses the standard error calculated using the sample proportion rather than the null value.

4.3.5 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

4.4 Module 3 and 4 Lab: Mixed Breed Dogs in the U.S.

4.4.1 Learning outcomes

- Determine whether simulation or theory-based methods of inference can be used.
- Analyze and interpret a study involving a single categorical variable.

4.4.2 Mixed Breed Dogs in the U.S.

The American Veterinary Medical Association estimated in 2010 that approximately 49% of dog owners in the U.S. own dogs that are classified as “mixed breed.” As part of a larger 2022 international study (Banton 2022) about overall dog health, survey participants were asked, among other things, to report whether their dog was purebred or a mixed breed. Seven hundred and fifty (750) dog owners from the U.S. were recruited to complete an online survey via an email indicating they had been randomly selected by Qualtrics (an “experience management” company that specializes in surveys). Three hundred sixty-four (364) out of 675 respondents from the U.S. reported they owned a mixed breed dog. Is there evidence that, in the last decade, the proportion of dog owners in the U.S. that own a mixed breed dog has changed from the value reported in 2010?

- Observational units:
- Variable:
- Type of variable:
- Success:

Activity intro

- Download the R script file and the data file (US_dogs.csv) from Canvas
- Upload both files to Canvas and open the R script file
- Enter the name of the dataset for datasetname.csv.
- Highlight and run lines 1 - 6

1. What is the value of the point estimate?

2. Create a plot of the data using the R code. Make sure to include an appropriate title with type of plot, observational units, and variable.

```
dogs %>% # Data set piped into...
  ggplot(aes(x = variable)) + # This specifies the variable
  geom_bar(aes(y = after_stat(prop), group = 1)) + # Tell it to make a bar plot with proportions
  labs(title = "Don't forget to title your plot",
        # Give your plot a title
        x = "Breed of Dog", # Label the x axis
        y = "Relative Frequency") # Label the y axis
```

Use statistical analysis methods to draw inferences from the data

3. Write out the parameter of interest in words, in context of the study.

4. Write out the null and alternative hypotheses in notation.

H_0 :

H_A :

5. Will theory-based methods give the same results as simulation based methods? Explain your answer.

Null Distribution

To use the computer simulation, we will need to enter the

- assumed “probability of success” (π_0),
- “sample size” (the number of observational units or cases in the sample),
- “number of repetitions” (the number of samples to be generated),
- “as extreme as” (the observed statistic), and
- the “direction” (matches the direction of the alternative hypothesis).

We will use the `one_proportion_test()` function in R (in the `catstats` package) to simulate the null distribution of sample proportions and compute a p-value.

- Using the provided R script file, fill in the values/words for each `xx` in the one proportion test to create a null distribution with 10000 simulations.
- Then highlight and run lines 21–26.

```
one_proportion_test(probability_success = xx, # Null hypothesis value
  sample_size = xx, # Enter sample size
  number_repetitions = 10000, # Enter number of simulations
  as_extreme_as = xx, # Observed statistic
  direction = "xx", # Specify direction of alternative hypothesis
  summary_measure = "proportion") # Reporting proportion or number of successes?
```

6. Report the p-value from the study.

Bootstrap distribution

We will use the `one_proportion_bootstrap_CI()` function in R (in the `catstats` package) to simulate the bootstrap distribution of sample proportions and calculate a confidence interval. Using the provided R script file, fill in the values/words for each `xx` in the one proportion bootstrap confidence interval (CI) code to create a bootstrap distribution with 10000 simulations. Then highlight and run lines 31–34 to create a 90% confidence interval.

```
one_proportion_bootstrap_CI(sample_size = xx, # Sample size
  number_successes = xx, # Observed number of successes
```

```
number_repetitions = 10000, # Number of bootstrap samples to use
confidence_level = xx) # Confidence level as a decimal
```

7. Report the 90% confidence interval.

Summarize the results of the study

8. Write a paragraph summarizing the results of the study. Be sure to describe:

- Summary statistic and interpretation
 - Summary measure (in context)
 - Value of the statistic
 - Order of subtraction when comparing two groups
- P-value and interpretation
 - Statement about probability or proportion of samples
 - Statistic (summary measure and value)
 - Direction of the alternative
 - Null hypothesis (in context)
- Confidence interval and interpretation
 - How confident you are (e.g., 90%, 95%, 98%, 99%)
 - Parameter of interest
 - Calculated interval
 - Order of subtraction when comparing two groups
- Conclusion (written to answer the research question)
 - Amount of evidence
 - Parameter of interest
 - Direction of the alternative hypothesis
- Scope of inference
 - To what group of observational units do the results apply (target population or observational units similar to the sample)?
 - What type of inference is appropriate (causal or non-causal)?

Upload a copy of your group's paragraph to Gradescope.

Paragraph (continued):

Unit 1 Review

The following module contains both a list of key topics covered in Unit 1 as well as Module Review Worksheets that will be covered in Weekly Review Sessions.

5.0.1 Key Topics

Review the key topics for Unit 1 prior to the first exams. All of these topics will be covered in Modules 1–4.

5.0.2 Module Review

The following worksheets review each of the modules. These worksheets will be completed during Melinda's Study Sessions each week. Solutions will be posted on Canvas in the Unit 1 Review folder after the study sessions.

5.1 Key Topics Exam 1

Descriptive statistics and study design

1. Identify the observational units.
2. Identify the types of variables (categorical or quantitative).
3. Identify the explanatory variable (if present) and the response variable (roles of variables).
4. Identify the appropriate type of graph and summary measure.
5. Identify if a given value is a statistic or a parameter. Identify the appropriate notation.
6. Identify the study design (observational study or randomized experiment).
7. Identify the sampling method and potential types of sampling bias (non-response, response, selection).
8. Identify and interpret the summary statistic
9. Identify the target population
10. Identify the types of sampling bias (response, non-response, selection, none)
11. Identify the type(s) of graph(s) that could be used to plot the given variable(s).

Hypothesis testing

12. Write the parameter of interest in context of the problem.
13. State the null and alternative hypotheses in both words and notation
14. Verify the validity condition is met to use simulation-based methods to find a p-value.
15. Verify the validity conditions are met to use theory-based methods to find a p-value from the theoretical distribution.
16. In a simulation-based hypothesis test, describe how to create one dot on a dotplot of the null distribution using coins, cards, or spinners.
17. Explain where the null distribution is centered and why.
18. Describe and illustrate how R calculates the p-value for a simulation-based test.
19. Describe and illustrate how R calculates the p-value for a theory-based test.
20. Type of theoretical distribution (standard normal distribution or t-distribution with appropriate degrees of freedom) used to model the standardized statistic in a theory-based hypothesis test.
21. Calculate and interpret the standard error of the statistic under the null using the correct formula on the Golden ticket.
22. Calculate and interpret the appropriate standardized statistic using the correct formula on the Golden ticket.
23. Interpret the p-value in context of the study: it is the probability of _____, assuming _____.
24. Evaluate the p-value for strength of evidence against the null: how much evidence does the p-value provide against the null?
25. Write a conclusion about the research question based on the p-value.
26. Describe which features of the study impact the p-value and how.

5.1.1 Confidence intervals

27. Describe how to simulate one bootstrapped sample using cards.
28. Explain where the bootstrap distribution is centered and why.
29. Find an appropriate percentile confidence interval using a bootstrap distribution from R output.
30. Verify the validity condition is met to use simulation-based methods to find the confidence interval.
31. Verify the validity conditions are met to use theory-based methods to calculate a confidence interval.
32. Describe and illustrate how the bootstrap distribution is used to find the confidence interval for a given confidence level.
33. Describe and illustrate how the standard normal distribution or t-distribution is used to find the multiplier for a given confidence level.
34. Calculate and interpret the standard error of the statistic (not assuming the null hypothesis) using the correct formula on the Golden ticket
35. Calculate the appropriate margin of error and confidence interval using theory-based methods.
36. Interpret the confidence interval in context of the study.
37. Based on the interval, what decision can you make about the null hypothesis? Does the confidence interval agree with the results of the hypothesis test? Justify your answer.
38. Interpret the confidence level in context of the study. What does “confidence” mean?
39. Describe which features of the study have an effect on the width of the confidence interval and how.

5.1.2 Probability

40. Calculate probabilities from a given table and give appropriate probability notation for both conditional and unconditional probabilities.
41. Create a two-way table using given probabilities.
42. Interpret a probability value in context of the problem.

5.2 Module 1 Review - Sampling Methods

1. Suppose that the proportion of all American adults that fit the medical definition of being obese is 0.23. A large medical clinic would like to determine if the proportion of their patients that are obese is higher than that of all American adults. The clinic takes a simple random sample of 30 of their patients and finds that 9 patients in the sample are obese.
 - a. What is the target population?
 - b. What are the observational units?
 - c. What variable is being studied?
 - d. Is the variable identified in part (c) categorical or quantitative?
2. Martha works in Macy's advertising department. She is interested in the shopping experience of all Macy's shoppers in the U.S. Every Saturday morning for a month she stands outside of the Bozeman Macy's asking people about their experience. One of the questions she uses is: "As a huge fan of Macy's, I believe Macy's has the best choices of clothing in Bozeman. Don't you agree?" Every person that was asked, responded.
 - a. Identify the target population.
 - b. Identify the sample.
 - c. Which of the three types of sampling bias (selection, non-response, response) may be present? Explain your choice(s).

3. This study aims to explore whether Swiss university students feeling academic study pressure (whether the student had experienced academic failure) tend to use psychotropic drugs (whether the student had used psychotropic drugs during the student's time at university) as a coping mechanism. An invitation email was sent to all bachelor's and master's students at the University of Lausanne, totaling 15,400 individuals, with a link to access the online questionnaire containing 49 questions and 107 items. No reminder was sent out, and no incentive was given to complete the questionnaire. A total of 1,690 students initially participated in the study, but 424 questionnaires were too incomplete to be used for analysis and were excluded. Additionally, 67 questionnaires were removed because of significant missing sociodemographic information, resulting in 1,199 completed responses included in the final analysis. Is there an association between study pressure and use of psychotropic drugs among Swiss University students?
- Identify the target population.
 - Identify the sample.
 - Which of the three types of sampling bias (selection, non-response, response) may be present? Explain your choice(s).
 - Identify the type and roles of each variable in the study.
4. Researchers decided to investigate whether a cat's coat color is associated with aggressive cat behavior by creating a 20-minute survey. The survey was distributed by posting it to social media and through cat-related listservs (e.g., For the Love of Cats), inviting individuals to take the survey. A total of 1,365 surveys were completed by participants. The frequency of each of the following aggressive behavior categories was assessed: hiss, stalk/chase, bite, slap/scratch. Frequency of behaviors toward people were recorded on a 6-point scale: 0 = never, 1 = less than once every 6 months, 2 = once every 6 months, 3 = once per month, 4 = once per week, 5 = one or more times per day. Because there were four aggressive behavior categories, each with a frequency of 0 to 5 possible, each cat could score between 0 to 20 for human aggression. Is there an association between coat color and aggressive behavior among cats?
- Identify the target population.
 - Identify the sample.
 - Which of the three types of sampling bias (selection, non-response, response) may be present? Explain your choice(s).
 - Identify the type and roles of each variable in the study.

5.3 Module 2 Review - Probability

1. Spelling errors in a text can either be non-word errors (teh instead of the) or word errors (lose instead of loose). It was found that non-word errors make up about 25% of all errors. A human proofreader will catch 92% of non-word errors and 75% of word errors.

Let N represent non-word errors and C represent that a human proofreader will catch the error.

- a. Identify the following values with appropriate probability notation.

0.25

0.92

0.75

- b. Fill in the table below to represent the situation:

	N	N^C	Total
C			
C^C			
Total			100000

- c. Using your table calculate the probability that a randomly selected error caught by a human proofreader is a non-word error. Use appropriate probability notation.

- d. Find the probability a selected error is a non-word error and was not caught by a human proofreader. Use appropriate probability notation.

- e. Find the value of $P(N|C)$. What does this probability mean?

2. A private college report contains these statistics:

- 70% of incoming freshmen attended public schools
- 75% of public-school students who enroll as freshmen eventually graduate
- 90% of other freshmen eventually graduate

Let A represent the event that a freshman attended public school and B the event that a freshman eventually graduates.

a. Identify the following values with appropriate probability notation.

0.70

0.75

0.90

b. Fill in the table below to represent the situation:

	A	A^C	Total
B			
B^C			
Total			100000

c. Calculate the probability a selected freshman attended public school given they did not graduate. Use appropriate probability notation.

d. Calculate the probability a selected freshman does not graduate. Use appropriate probability notation.

e. Of the population of freshman that attended public school, what is the probability they do not graduate. Use appropriate probability notation.

f. Find the value of $P(A \text{ and } B^C)$. Write this probability in context of the problem.

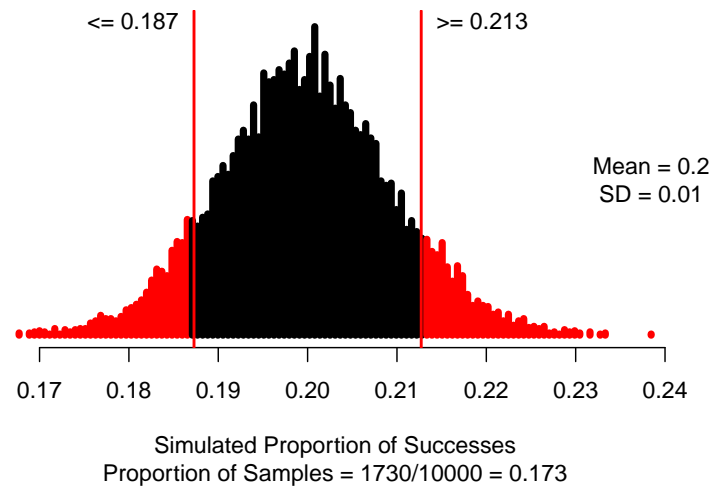
5.4 Module 3 Review - Simulation Methods for a Single Proportion

```
hearing <- read.csv("data/hearing_loss.csv")
```

A recent study examined hearing loss data for 1753 U.S. teenagers. In this sample, 328 were found to have some level of hearing loss. News of this study spread quickly, with many news articles blaming the prevalence of hearing loss on the higher use of ear buds by teens. At MSNBC.com (8/17/2010), Carla Johnson summarized the study with the headline: “1 in 5 U.S. teens has hearing loss, study says.” Is this an appropriate or a misleading headline?

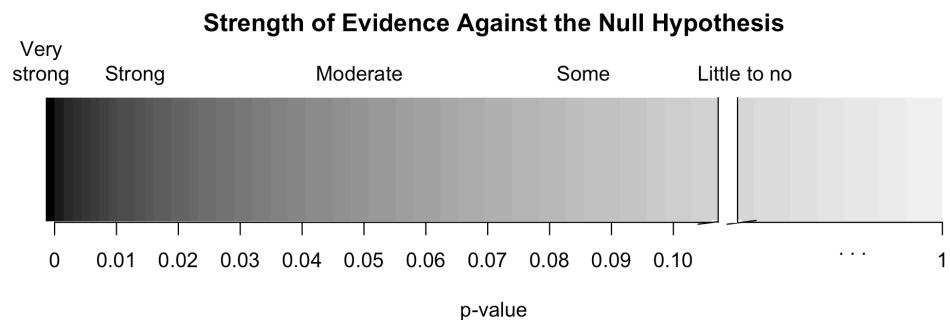
1. Write the parameter of interest in context of the study.
2. Write the null hypothesis in words and notation in context of the problem.
3. Based on the research questions, choose the direction for the alternative hypothesis.
4. Write the alternative hypothesis in words and notation in context of the problem.
5. Calculate the summary statistic. Use proper notation.
6. What values should be entered for each of the following into the one proportion test to create 10000 simulations?
 - Probability of success:
 - Sample size:
 - Number of repetitions:
 - As extreme as:
 - Direction (“greater”, “less”, or “two-sided”):

```
one_proportion_test(probability_success = 0.2, #Null hypothesis value
  sample_size = 1753, #Enter sample size
  number_repetitions = 10000, #Enter number of simulations
  as_extreme_as = 0.187, #observed statistic
  direction = "two-sided", #specify direction of alternative hypothesis
  summary_measure = "proportion") #Reporting proportion or number of successes?
```



7. Interpret the p-value in context of the problem.

8. How much evidence does the data provide against the null hypothesis?



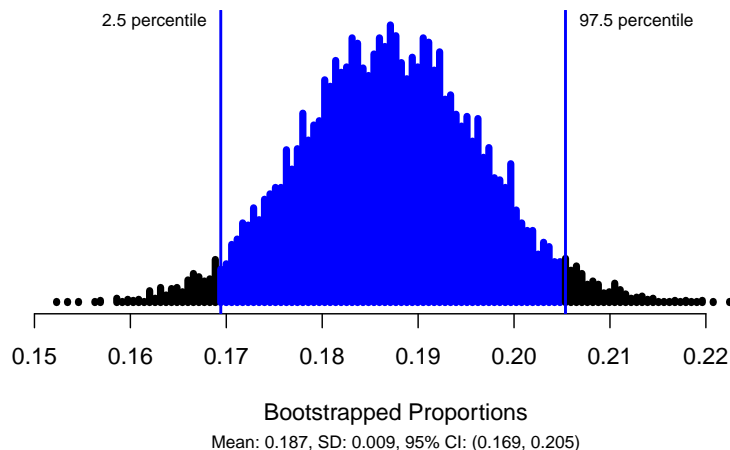
9. Write a conclusion to the study in context of the problem.

10. Would a 95% confidence interval contain the null value of 0.2? Explain.

11. What values should be entered for each of the following into the simulation to create the bootstrap distribution of sample proportions to find a 95% confidence interval?

- Sample size:
- Number of successes:
- Number of repetitions:
- Confidence level (as a decimal):

```
set.seed(216)
one_proportion_bootstrap_CI(sample_size = 1753, # Sample size
                             number_successes = 328, # Observed number of successes
                             number_repetitions = 10000, # Number of bootstrap samples to use
                             confidence_level = 0.95) # Confidence level as a decimal
```



12. Explain how to use cards to create one bootstrap sample.

13. Report the 95% confidence interval in interval notation.

14. Interpret the 95% confidence interval in context of the problem.

5.5 Module 4 Review - Theory-based Methods for a Single Proportion

Statistician Jessica Utts has conducted an extensive analysis of Ganzfeld studies that have investigated psychic functioning. Ganzfeld studies involve a “sender” and a “receiver.” Two people are placed in separate rooms. The sender looks at a “target” image on a television screen and attempts to transmit information about the target to the receiver. The receiver is then shown four possible choices or targets, one of which is the correct target and the other three are “decoys.” The receiver must choose the one he or she thinks best matches the description transmitted by the sender. If the correct target is chosen by the receiver, the session is a “hit.” Otherwise, it is a miss. Utts reported that her analysis considered a total of 2,124 sessions and found a total of 709 “hits” (Utts, 2010). Is there evidence of psychic ability?

1. Write the parameter of interest in context of the study.
2. Calculate the point estimate. Use proper notation.
3. Write the null hypothesis in words.
4. Write the alternative hypothesis in notation.

A single proportion can be mathematically modeled using the normal distribution if certain conditions are met. Conditions for the sample distribution of \hat{p} .

- Independence: The sample’s observations are independent, e.g., are from a simple random sample
- Large enough sample size:
 - Success-Failure Condition: There are at least 10 successes and 10 failures in the sample

$$n \times \hat{p} \geq 10$$

and

$$n \times (1 - \hat{p}) \geq 10$$

5. Are the conditions met to model the data with the Normal distribution?

Standardized sample proportion.

The standardized statistic for theory-based methods for one proportion is:

$$Z = \frac{\hat{p} - \pi_0}{SE_0(\hat{p})}$$

Where

$$SE_0(\hat{p}) = \sqrt{\frac{\pi_0 \times (1 - \pi_0)}{n}}$$

6. Calculate the null standard error of the sample proportion

7. Calculate the standardized statistic for the sample proportion.

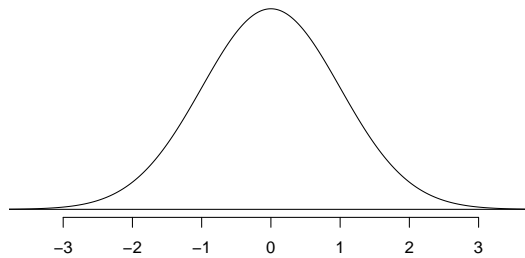
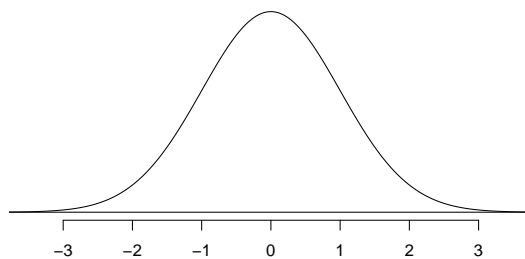


Figure 5.1: A standard normal curve.



8. Interpret the standardized statistic in context of the problem.

We will use the `pnorm()` function in R to find the p-value. The value of the standardized statistic calculated in question 8 is entered into the R code. We used `lower.tail = FALSE` to find the p-value so that R will calculate the p-value *greater* than the value of the standardized statistic.

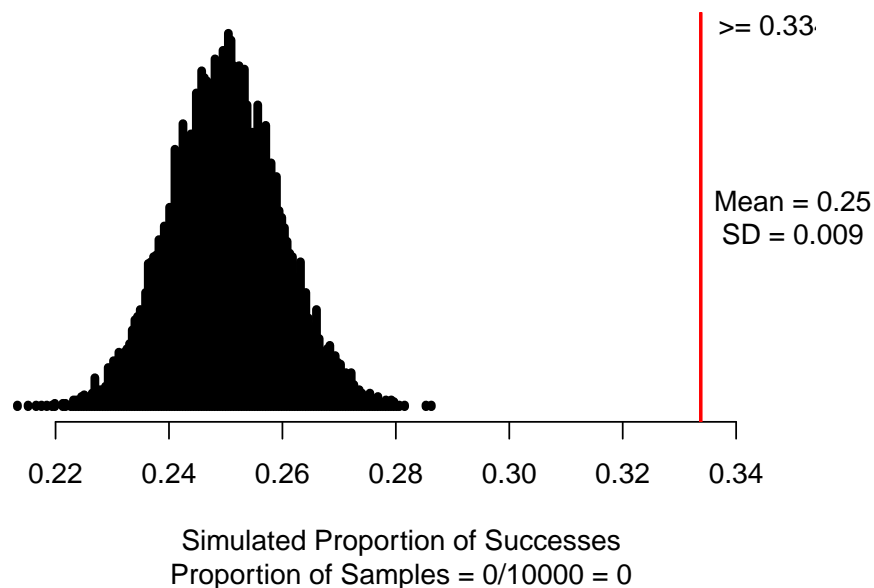
Notes:

- Use `lower.tail = TRUE` when doing a left-sided test.
- Use `lower.tail = FALSE` when doing a right-sided test.
- To find a two-sided p-value, use a left-sided test for negative Z or a right-sided test for positive Z, then multiply the value found by 2 to get the p-value.

```
pnorm(9.333, # Enter value of standardized statistic
      m=0, s=1, # Using the standard normal mean = 0, sd = 1
      lower.tail=FALSE) # Gives a p-value greater than the standardized statistic
#> [1] 5.145792e-21
```

9. Report the value of the p-value.

Simulation Method:



10. Interpret the p-value in context of the study.

Next we will use theory-based methods to estimate the parameter of interest.

To calculate a theory-based 95% confidence interval for π , we will first find the **standard error** of \hat{p} by plugging in the value of \hat{p} for π in $SD(\hat{p})$:

$$SE(\hat{p}) = \sqrt{\frac{\hat{p} \times (1 - \hat{p})}{n}}.$$

Note that we do not include a “0” subscript, since we are not assuming a null hypothesis.

11. Calculate the standard error of the sample proportion to find a 95% confidence interval.

To find the confidence interval, we will add and subtract the **margin of error** to the point estimate:

point estimate \pm margin of error

$$\hat{p} \pm z^* SE(\hat{p})$$

The z^* multiplier is the percentile of a standard normal distribution that corresponds to our confidence level. If our confidence level is 95%, we find the Z values that encompass the middle 95% of the standard normal distribution. If 95% of the standard normal distribution should be in the middle, that leaves 5% in the tails, or 2.5% in each tail. The `qnorm()` function in R will tell us the z^* value for the desired percentile (in this case, 95% + 2.5% = 97.5% percentile).

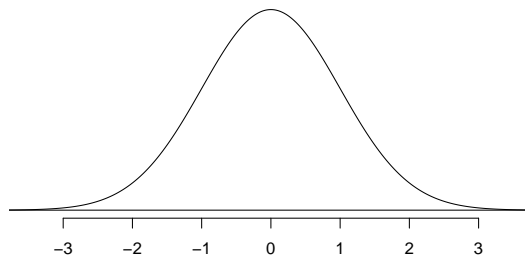


Figure 5.2: A standard normal curve.

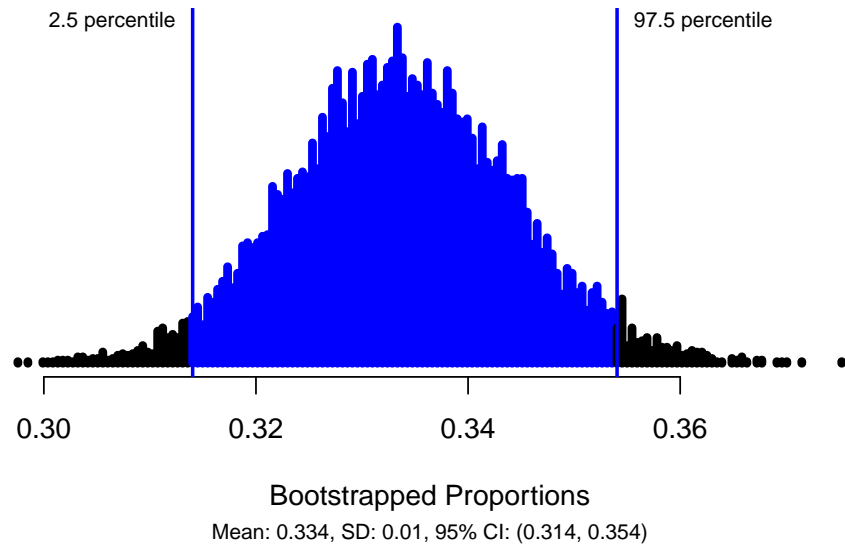
```
qnorm(0.975) # Multiplier for 95% confidence interval
```

```
#> [1] 1.959964
```

12. Calculate the margin of error for a 95% confidence interval for the true proportion of sessions that will result in a hit.

13. Calculate the 95% confidence interval for the true proportion of sessions that will result in a hit.

Simulation Methods:



14. Interpret the 95% confidence interval in context of the problem.
15. Write a conclusion based on the p-value and the 95% confidence interval.

5.6 Group Exam 1 Review

Use the provided data set from the Islands (Bulmer, n.d.) (Exam1ReviewData.csv) and the appropriate Exam 1 Review R script file to answer the following questions. Each adult (>21) islander was selected at random from all adult islanders. Note that some islanders choose not to participate in the study. These islanders that did not consent to be in the study are removed from the dataset before analysis. Variables and their descriptions are listed below. Here is some more information about some of the variables collected. Music type (classical or heavy metal) was randomly assigned to the Islanders. Time to complete the puzzle cube was measured after listening to music for each Islander. Heart rate and blood glucose levels were both measured before and then after drinking a caffeinated beverage.

Variable	Description
Island	Name of Island that the Islander resides on
City	Name of City in which the Islander resides
Population	Population of the City
Name	Name of Islander
Consent	Whether the Islander consented to be in the study (Declined , Consented)
Gender	Gender of Islander (M = male, F = Female)
Age	Age of Islander
Married	Marital status of Islander (yes , no)
Smoking_Status	Whether the Islander is a current smoker (nonsmoker , smoker)
Children	Whether the Islander has children (yes , no)
weight_kg	Weight measured in kg
height_cm	Height measured in cm
respiratory_rate	Breaths per minute
Type_of_Music	Music type Islander was randomly assigned to listen to (Classical , Heavy Metal)
After_PuzzleCube	Time to complete puzzle cube (minutes) after listening to assigned music
Education_Level	Highest level of education completed (highschool , university)
Balance_Test	Time balanced measured in seconds with eyes closed
Blood_Glucose_before	Level of blood glucose (mg/dL) before consuming assigned drink
Heart_Rate_before	Heart rate (bpm) before consuming assigned drink
Blood_Glucose_after	Level of blood glucose (mg/dL) after consuming assigned drink
Heart_Rate_after	Heart rate (bpm) after consuming assigned drink
Diff_Heart_Rate	Difference in heart rate (bpm) for Before - After consuming assigned drink
Diff_Blood_Glucose	Difference in blood glucose (mg/dL) for Before - After consuming assigned drink

1. What are the observational units?
2. In the table above, indicate which variables are categorical (C) and which variables are quantitative (Q).
3. What type of bias may be present in this study? Explain.

4. Use the appropriate Exam 1 Review R script file to find the summary statistic and graphical display of the data to assess the following research question, “Is there evidence that the proportion of adult Islanders that smoke differs from the reported value of 11%?”
- a. What is the name of the variable to be assessed in this research question?

What type of variable (categorical or quantitative) is the variable you identified?

- b. Use the R script file to get the counts for each level of the variable. Fill in the following table with the variable name, levels of the variable, and counts using the values from the R output.

	Count
Success	
Failure	
Total	

- c. Calculate the value of the summary statistic to answer the research question. Give appropriate notation.
- d. What type of graph(s) would be appropriate for this research question?
- e. Using the provided R file create a graph of the data. Sketch the graph below:

f. Assess if the following conditions are met:

Independence (needed for both simulation and theory-based methods):

Success-Failure (must be met to use theory-based methods):

g. Use the provided R script file to find the simulation p-value to assess the research question. Report the p-value.

h. Interpret the p-value in the context of the problem.

i. Write a conclusion to the research question based on the p-value.

j. Using a significance level of $\alpha = 0.05$, what statistical decision will you make about the null hypothesis?

k. Use the provided R script file to find a 95% confidence interval.

l. Interpret the 95% confidence interval in context of the problem.

m. Regardless to your answer in part f, calculate the standardized statistic.

n. Interpret the value of the standardized statistic in context of the problem.

- o. Use the provided R script file to find the theory-based p-value.
- p. Use the provided R script file to find the appropriate z^* multiplier and calculate the theory-based confidence interval.
- q. Does the theory-based p-value and CI match those found using simulation methods? Explain why or why not.
- r. To what group of observational units do the results apply?

Exploring Quantitative Data: Exploratory Data Analysis and Hypothesis Testing for a Single Quantitative Variable

6.1 Vocabulary Review and Key Topics

Review the Golden Ticket posted in the resources at the end of the coursepack for a summary of a single quantitative variable.

6.1.1 Key topics

Module 6 will introduce exploratory data analysis and hypothesis testing using both simulation-based and theory-based methods for a single quantitative variable. The **summary measure** for one quantitative variable is the **mean**. Additionally, we can find the five number summary (min, Q1, median, Q3, max) as well as the sample standard deviation.

- Notation for a sample mean: \bar{x}
- Notation for a sample standard deviation: s
- Notation for a population mean: μ
- Notation for a population standard deviation: σ
- Types of plots for a single categorical variable:
 - Histogram
 - Boxplot
 - Dotplot

6.1.2 Vocabulary

Sample statistics for a single quantitative variable

- **Mean**, \bar{x} : the average

$$\bar{x} = \frac{x_1 + x_2 + \cdots + x_n}{n},$$

where x_1, x_2, \dots, x_n are the data values and n is the sample size.

- **Median**: value at the 50th percentile; approximately 50% of data values are at or below the value of the median.
- **Quartile 1** (lower quartile), Q_1 : value at the 25th percentile; approximately 25% of data values are at or below the value of Q_1 .
- **Quartile 3** (upper quartile), Q_3 : value at the 75th percentile; approximately 75% of data values are at or below the value of Q_3 .

- **Sample standard deviation**, s : on average, each value in the data set is s units from the mean of the data set (\bar{x}). We will always calculate s using R, but it is calculated using the following formula:

$$s = \sqrt{\frac{(x_1 - \bar{x})^2 + (x_2 - \bar{x})^2 + \dots + (x_n - \bar{x})^2}{n - 1}},$$

where x_1, x_2, \dots, x_n are the data values, \bar{x} is the sample mean, and n is the sample size.

- **Interquartile range**: the range of the data between the two quartiles: $IQR = Q_3 - Q_1$.
- R code to find the summary statistics for a quantitative variable:

```
object %>% # Data set piped into...
  summarise(favstats(variable))
```

Plotting one quantitative variable

- **Histogram**: sorts a quantitative variable into bins of a certain width. R code to create a histogram:

```
object %>% # Data set piped into...
  ggplot(aes(x = variable)) + # Name variable to plot
  geom_histogram(binwidth = 10) + # Create histogram with specified binwidth
  labs(title = "Don't forget to title the plot!", # Title for plot
       x = "x-axis label", # Label for x axis
       y = "y-axis label") # Label for y axis
```

- **Boxplot**: plots the values of the five-number summary and shows any outliers in the data set. R code to create a boxplot:

```
object %>% # Data set piped into...
  ggplot(aes(x = variable)) + # Name variable to plot
  geom_boxplot() + # Create boxplot
  labs(title = "Don't forget to title the plot!", # Title for plot
       x = "x-axis label", # Label for x axis
       y = "y-axis label") # Label for y axis
```

- **Dotplot**: plots each value as a dot along the x -axis. R code to create a dotplot:

```
object %>% # Data set piped into...
  ggplot(aes(x = variable)) + # Name variable to plot
  geom_dotplot() + # Create dotplot
  labs(title = "Don't forget to title the plot!", # Title for plot
       x = "x-axis label", # Label for x axis
       y = "y-axis label") # Label for y axis
```

- Four characteristics of a distribution of a single quantitative variable:
 - Shape (symmetric, skewed left, or skewed right)
 - Center
 - Spread
 - Outliers?

Hypothesis testing for a single mean

- **Hypotheses in notation for a single mean:** In the hypotheses below, μ_0 is the **null value**.

$$H_0 : \mu = \mu_0$$
$$H_A : \mu \left\{ \begin{array}{c} < \\ \neq \\ < \end{array} \right\} \mu_0$$

Simulation-based hypothesis testing

- **Conditions necessary to use simulation-based methods for inference for a single quantitative variable:**
 - **Independence:** observational units must be independent of one another.
- **Simulation-based methods to create the null distribution:** R code to use for simulation-based methods for one quantitative variable to find the p-value, `one_mean_test` (from the `catstats` package), is shown below. Review the comments (instructions after the `#`) to see what each should be entered for each line of code.

```
one_mean_test(object$variable, #Enter the object name and variable
  null_value = xx, #Enter the null value for the study
  summary_measure = "mean", #Can choose between mean or median
  shift = xx, #Difference between the null value and the sample mean
  as_extreme_as = xx, #Value of the summary statistic
  direction = "xx", #Specify direction of alternative hypothesis
  number_repetitions = 10000)
```

Theory-based hypothesis testing

- Theory-based methods should give the same results as simulation-based methods if conditions are met. For a single quantitative variable, conditions are met if either the data themselves follow a normal distribution or if the sample size is large enough. We call this the “normality condition.”
- **Conditions for the sampling distribution of \bar{x} to follow an approximate normal distribution:**
 - **Independence:** the sample’s observations are independent, e.g., are from a simple random sample. (*Remember:* This also must be true to use simulation methods!)
 - **Normality Condition:** either the sample observations come from a normally distributed population or we have a large enough sample size. To check this condition, use the following rules of thumb:
 - * $n < 30$: The distribution of the sample must be approximately normal with no outliers.
 - * $30 \leq n < 100$: We can relax the condition a little; the distribution of the sample must have no extreme outliers or skewness.
 - * $n \geq 100$: Can assume the sampling distribution of \bar{x} is nearly normal, even if the underlying distribution of individual observations is not.
- **t-distribution:** a theoretical distribution that is bell-shaped with mean zero. Its degrees of freedom determine the variability of the distribution. For very large degrees of freedom, the t -distribution is close to a standard normal distribution. For a single quantitative variable, the degrees of freedom are calculated by subtracting one from the sample size: $n - 1$. A t -distribution with $n - 1$ degrees of freedom is denoted by: t_{n-1} .

- **Standard error of the sample mean:** measures the how far each possible sample mean is from the true mean, on average, and is calculated using the formula below:

$$SE(\bar{x}) = \frac{s}{\sqrt{n}}$$

where s is the sample standard deviation.

- For inference involving means, the formula for the standard error will be the same for both hypothesis tests and confidence intervals (unlike inference involving proportions, where the standard error for a hypothesis test used the null value in the calculation).

- **Standardized sample mean:** standardized statistic for a single quantitative variable calculated using:

$$T = \frac{\bar{x} - \mu_0}{SE(\bar{x})},$$

If the conditions for the sampling distribution of \bar{x} to follow an approximate normal distribution are met, and if the true value of μ is equal to the null value of μ_0 , the standardized sample mean, T , will have an approximate t -distribution with $n - 1$ degrees of freedom.

- The theory-based **p-value** for hypothesis testing involving means can be found in R by using the **pt** function to find the probability of the observed standardized statistic or one more extreme (in the direction of H_A). This probability is the area under a *t-distribution with the appropriate degrees of freedom* at or more extreme than the observed standardized statistic.

- **pt** will give you a p-value using the t -distribution with a given degrees of freedom (enter for **yy**). For a single mean, **df** = $n - 1$.
- Enter the value of the standardized statistic for **xx**
- If a “greater than” alternative, change **lower.tail** = TRUE to FALSE.
- If a two-sided test, multiply by 2.

```
pt(xx, df = yy, lower.tail=TRUE)
```

6.2 Activity 9: Summarizing Quantitative Variables

6.2.1 Learning outcomes

- Identify and create appropriate summary statistics and plots given a data set or research question for quantitative data.
- Interpret the following summary statistics in context: median, lower quartile, upper quartile, standard deviation, interquartile range.

6.2.2 Terminology review

In today's activity, we will review summary measures and plots for quantitative variables. Some terms covered in this activity are:

- Two measures of center: mean, median
- Two measures of spread (variability): standard deviation, interquartile range (IQR)
- Plots of quantitative variables: dotplots, boxplots, histograms
- Given a plot or set of plots, describe and compare the distribution(s) of quantitative variables (center, variability, shape, outliers).

To review these concepts, see Chapter 5 in the textbook.

6.2.3 The Integrated Postsecondary Education Data System (IPEDS)

These data were collected on a subset of higher education institutions that met the following selection criteria (Education Statistics 2018):

- Degree granting
- United States only
- Title IV participating
- Not for profit
- 2-year or 4-year or above
- Has full-time first-time undergraduates

Some of the variables collected and their descriptions are below. Note that several variables have missing values for some institutions (denoted by "NA").

Variable	Description
UnitID	Unique institution identifier
Name	Institution name
State	State abbreviation
Sector	whether public or private
LandGrant	Is this a land-grant institution (Yes/No)
Size	Institution size category based on total student enrolled for credit, Fall 2018: Under 1,000, 1,000\$-4,999, 5,000-9,999, 10,000-\$19,999, 20,000 and above
Cost_OutofState	Cost of attendance for full-time out-of-state undergraduate students
Cost_InState	Cost of attendance for full-time in-state undergraduate students
Retention	Retention rate is the percent of the undergraduate students that re-enroll in the next year
Graduation_Rate	6-year graduation rate for undergraduate students

Variable	Description
SATMath_75	75th percentile Math SAT score
ACT_75	75th percentile ACT score

Identifying variables in a data set

Look through the provided table of variable descriptions. The **UnitID** and **Name** are identifiers for each observational unit, *US degree-granting higher education institutions in 2018*.

1. Identify in the table which variables collected on the US institutions are categorical (C) and which variables are quantitative (Q).

Summarizing quantitative variables

Notes on Summarizing Quantitative Variables:

The `favstats()` function from the `mosaic` package gives the summary statistics for a quantitative variable. The R output below provides the summary statistics for the variable **Graduation_Rate**. The summary statistics provided are the two measures of center (mean and median) and two measures of spread (standard deviation and the quartile values to calculate the IQR) for undergraduate 6-year graduation rate.

- Highlight and run lines 1–12 in the provided R script file to load the data set. Check that the summary statistics match the output given in the coursepack.
- Notice that the 2-year institutions were removed so the observational units for this study are **4-year US degree-granting higher education institutions in 2018**.

```
IPEDS <- read.csv("https://www.math.montana.edu/courses/s216/data/IPEDS_2018.csv")
IPEDS <- IPEDS %>%
  filter(Sector != "Public 2-year") # Filters the data set to remove Public 2-year
IPEDS <- IPEDS %>%
  filter(Sector != "Private 2-year") # Filters the data set to remove Private 2-year
IPEDS %>%
  summarize(favstats(Graduation_Rate))
```

```
#>   min Q1 median Q3 max      mean      sd    n missing
#> 1    0  38     53  67 100 52.48749 20.63192 1918      49
```

Two measures of center:

- Mean:
- Median:

Two measures of spread:

- Standard deviation:
- Interpretation of the standard deviation:

- Interquartile range: $IQR = Q_3 - Q_1$:

- Interpretation of Q_3 :

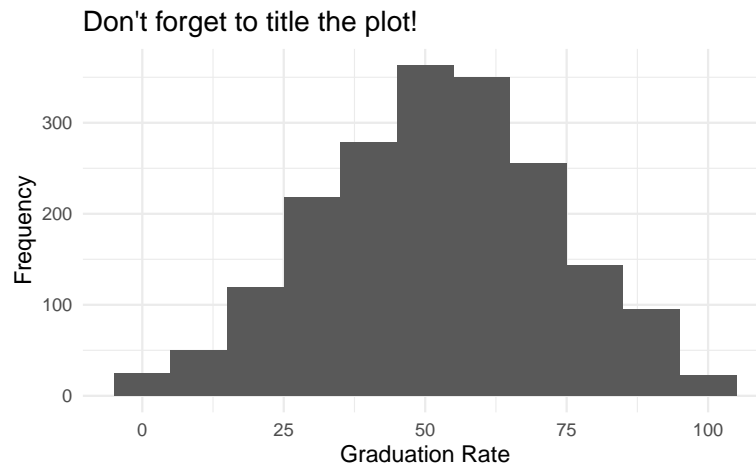
Displaying a single quantitative variable

There are three type of plots used to plot a single quantitative variable: a dotplot, a histogram or a boxplot. A dotplot of graduation rates would plot a dot for the graduation rate for each 4-year US higher education institution.

First, let's create a histogram of the variable `Graduation_Rate`.

- Enter the name of the variable, `Graduation_Rate` in line 19 for `variable` in the R script file.
- Replace the word title for the plot in line 21 between the quotations with a descriptive title. **A title should include: type of plot, variable or variables plotted, and observational units.**
- Highlight and run lines 18–23 to create the histogram.

```
IPEDS %>% # Data set piped into...
ggplot(aes(x = Graduation_Rate)) + # Name variable to plot
  geom_histogram(binwidth = 10) + # Create histogram with specified binwidth
  labs(title = "Don't forget to title the plot!", # Title for plot
        x = "Graduation Rate", # Label for x axis
        y = "Frequency") # Label for y axis
```



Notice that the **bin width** for the histogram is 10. For example the first bin consists of the number of institutions in the data set with a graduation rate of 0 to 10%. It is important to note that a graduation rate on the boundary of a bin will fall into the bin above it; for example, 20 would be counted in the bin 20–30.

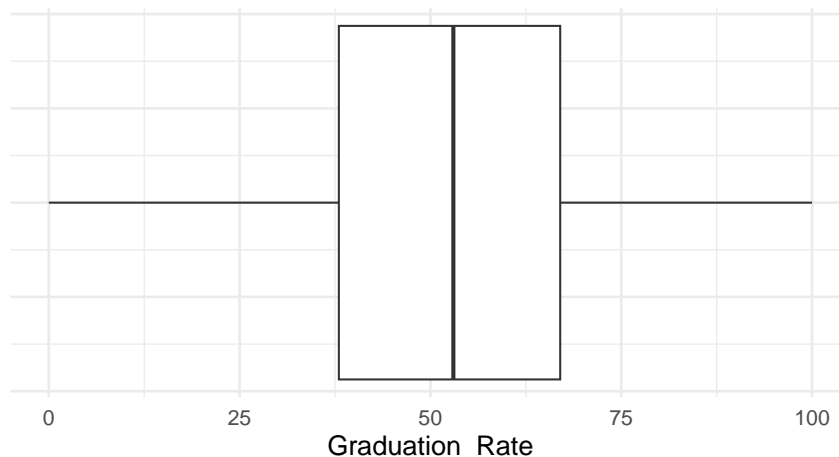
Which range of Graduation Rates have the highest frequency?

Next we will create a boxplot of the variable `Graduation_Rate`.

- Enter the name of the variable in line 28 for `variable` in the R script file.
- Highlight and run lines 28–36 to create the boxplot.

```
IPEDS %>% # Data set piped into...
ggplot(aes(x = Graduation_Rate)) + # Name variable to plot
  geom_boxplot() + # Create boxplot with specified binwidth
  labs(title = "Boxplot of Graduation Rates for \n 4-year Higher Education Institutions",
        # Title for plot
        # Note the \n starts a new line
        x = "Graduation_Rate", # Label for x axis
        y = "") + # Remove y axis label
  theme(axis.text.y = element_blank(),
        axis.ticks.y = element_blank()) # Removes y-axis ticks
```

Boxplot of Graduation Rates for 4-year Higher Education Institutions



Use the following formulas to find the invisible fence on both ends of the distribution. Draw a dotted line at the invisible fence to show how the outliers were detected (any values less than the lower fence or greater than the upper fence were flagged as outliers).

$$\text{Lower Fence: } Q_1 - 1.5 \times IQR \quad \text{Upper Fence: } Q_3 + 1.5 \times IQR$$

When describing distributions of quantitative variables we discuss the **shape** (symmetric or skewed), the **center** (mean or median), **spread** (standard deviation or IQR), and if there are **outliers** present.

2. What is the shape of the distribution of graduation rates?
3. From which plot (histogram or boxplot) is it easier to determine the shape of the distribution?
4. From which plot is it easier to determine if there are outliers?

Robust statistics

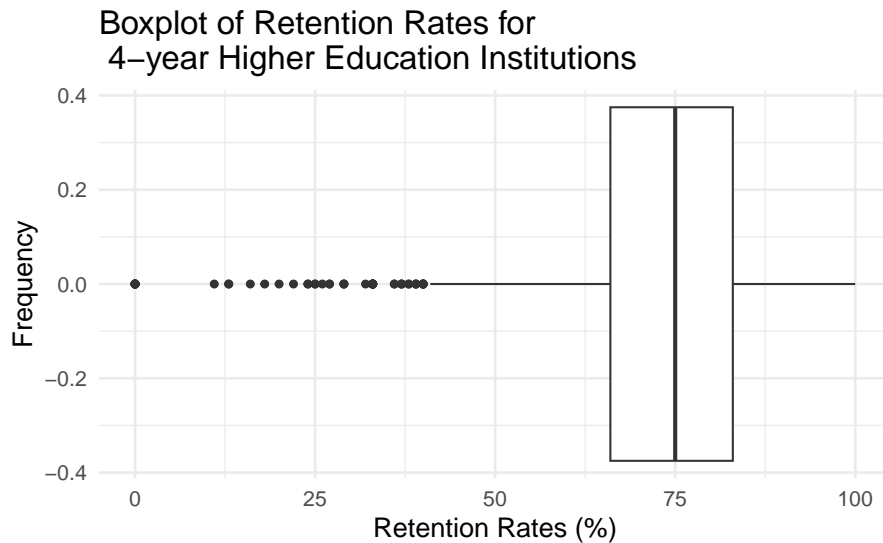
Let's examine how the presence of outliers affects the different summary measures for center and spread. For this part of the activity, we will look at the retention rate variable (**Retention**) in the IPEDS data set.

```
IPEDS %>% # Data set piped into...
  summarise(favstats(Retention))
#>   min Q1 median Q3 max   mean    sd   n missing
#> 1    0 66    75 83 100 73.8525 15.14323 1817      150

IPEDS %>% # Data set piped into...
  ggplot(aes(x = Retention)) + # Name variable to plot
  geom_boxplot() + # Create boxplot
  labs(title = "Boxplot of Retention Rates for \n 4-year Higher Education Institutions",
```



```
# Title for plot
x = "Retention Rates (%)", # Label for x axis
y = "Frequency" # Label for y axis
```



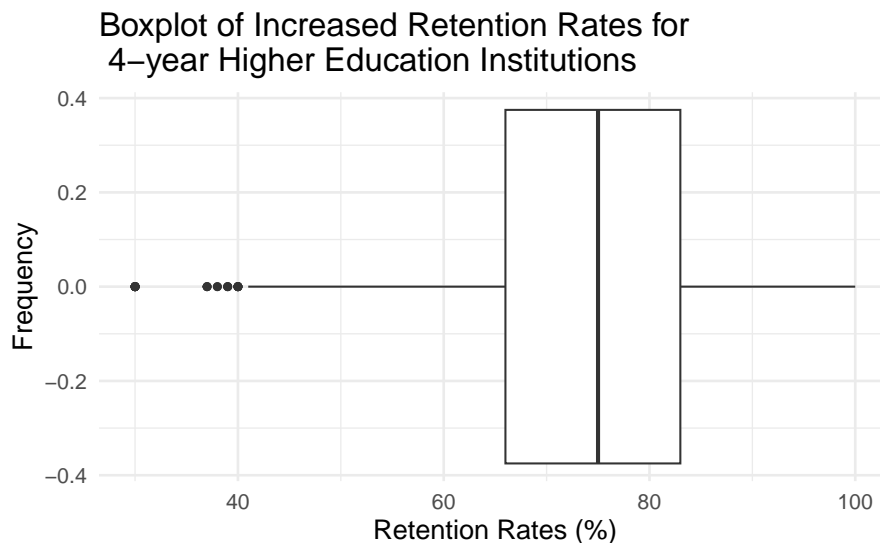
5. Report the values for the two measures of center for these data.

6. Report the values for the two measures of spread for these data.

To show the effect of outliers on the measures of center and spread, the smallest values of retention rate in the data set were increased by 30%. This variable is called `Retention_Inc`.

```
IPEDS %>% # Data set piped into...
  summarise(favstats(Retention_Inc))
#>   min Q1 median Q3 max    mean    sd   n missing
#> 1  30 66    75 83 100 74.49642 13.41255 1817    150

IPEDS %>% # Data set piped into...
  ggplot(aes(x = Retention_Inc)) + # Name variable to plot
  geom_boxplot() + # Create histogram
  labs(title = "Boxplot of Increased Retention Rates for \n 4-year Higher Education Institutions",
        # Title for plot
        x = "Retention Rates (%)", # Label for x axis
        y = "Frequency" # Label for y axis)
```



7. Report the values for the two measures of center for this new data set.
8. Report the values for the two measures of spread for this new data set.
9. Which measure of center is robust to outliers? Explain your answer.
10. Which measure of spread is robust to outliers? Explain your answer.

6.2.4 Take-home messages

1. Histograms, box plots, and dot plots can all be used to graphically display a single quantitative variable.
2. The box plot is created using the five number summary: minimum value, quartile 1, median, quartile 3, and maximum value. Whiskers extend to the lowest value and highest value that are *not* considered outliers. Values in the data set that are less than $Q_1 - 1.5 \times IQR$ or greater than $Q_3 + 1.5 \times IQR$ are considered outliers and are graphically represented by a dot outside of the whiskers on the box plot.
3. Data should be summarized numerically and displayed graphically to give us information about the study.
4. When comparing distributions of quantitative variables we look at the shape, center, spread, and for outliers. In this course, we only consider two measures of center (mean and the median), and two measures of spread (standard deviation and the interquartile range, $IQR = Q_3 - Q_1$).

6.2.5 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

6.3 Activity 10: Inference for a Single Quantitative Variable: Simulation Methods

6.3.1 Learning outcomes

- Given a research question involving one quantitative variable, construct the null and alternative hypotheses in words and using appropriate statistical symbols.
- Investigate the process of creating a null distribution for one quantitative variable.
- Find, evaluate, and interpret a p-value from the null distribution.

6.3.2 Terminology review

In today's activity, we will use simulation-based and theory-based methods to analyze a single quantitative variable. Some terms covered in this activity are:

- Null hypothesis
- Alternative hypothesis
- Null distribution
- Bootstrap distribution
- p-value

To review these concepts, see Chapters 9 and 17 in the textbook.

6.3.3 College student sleep habits

According to an article in *Sleep* (Watson 2015), experts recommend adults (>18 years old) get at least 7 hours of sleep per night. A professor at MSU is interested in the sleep habits of MSU students. The professor obtained a representative sample of MSU students and asked each student to report the amount of sleep they get on a typical night. Is there evidence that MSU students get less than the recommended 7 hours of sleep per night, on average?

- Observational units:
- Variable:
- Type of variable:

Summarizing quantitative variables

- Download the R script file and data file for this activity
- Upload both files to the RStudio server and open the R script file
- Enter the name of the dataset for datasetname.csv
- Highlight and run lines 1–8 to load the data

```
sleep <- read.csv("datasetname.csv")
```

Ask a research question

Parameter of interest in context of the study:

Null Hypothesis (in words):

Null Hypothesis (in notation):

Alternative Hypothesis (in words):

Alternative Hypothesis (in notation):

Summarize and visualize the data

The `favstats()` function from the `mosaic` package gives the summary statistics for a quantitative variable.

- Enter the variable name, `SleepHours`, for `variable` in line 13
- Highlight and run lines 12–13

```
sleep %>%  
  summarize(favstats(variable))
```

1. About how far is each number of hours of sleep for a Stat 216 student from the mean number of hours of sleep, on average?

Create a boxplot of the variable `SleepHours`.

- Enter the name of the variable in line 19 for `variable` in the R script file.
- Enter a title in line 21 for the plot between the quotations.
- Highlight and run lines 18–25.

```
sleep %>% # Data set piped into...  
  ggplot(aes(x = variable)) + # Name variable to plot  
  geom_boxplot() + # Create boxplot with specified binwidth  
  labs(title = "Don't forget to title your plot!", # Title for plot  
       x = "Amount of sleep (hrs)", # Label for x axis  
       y = "") + # Remove y axis label  
  theme(axis.text.y = element_blank(),  
        axis.ticks.y = element_blank()) # Removes y-axis ticks
```

2. Describe the distribution of number of hours of sleep using the four characteristics of boxplots.

Simulation methods

Notes on simulation methods for a single mean

To simulate the null distribution of sample means we will use a bootstrapping method. Recall that the null distribution must be created under the assumption that the null hypothesis is true. Therefore, before bootstrapping, we will need to *shift* each data point by the difference $\mu_0 - \bar{x}$. This will ensure that the mean of the shifted data is μ_0 (rather than the mean of the original data, \bar{x}), and that the simulated null distribution will be centered at the null value.

- Calculate the difference $\mu_0 - \bar{x}$. Based on the sign of this difference, will we need to shift the data up or down?

Your instructor will demonstrate how the shift is performed in Excel.

- Open the data set (`sleep_college`) in Excel.
- Create a new column labeled Shift.
- In the column, Shift, add the shifted value to each value in the `SleepHours` column.

```
sleep <- read.csv("sleep_college.csv")
sleep %>%
  summarize(favstats(Shift))
```

3. Report the mean of the `Shift` variable. Why does it make sense that this value is the same as the null value?
4. Report the standard deviation of the `Shift` variable. How does this compare to the standard deviation for the variable `SleepHours`? Explain why these values are the same.

5. What inputs should be entered for each of the following to create the simulated null distribution?

- Null value (What is the null value for the study?):
- Summary measure ("mean" or "median"):
- Shift (difference between $\mu_0 - \bar{x}$):
- As extreme as (enter the value for the observed sample mean):
- Direction ("greater", "less", or "two-sided"):
- Number of repetitions:

The `one_mean_test` will be used to find the p-value for the simulation test. Following the instructions below to complete the code.

- Enter your answers for question 5 in place of the `xx`'s to produce the null distribution with 10000 simulations.
- Highlight and run lines 36–42.

```
one_mean_test(sleep$SleepHours, #Enter the object name and variable
               null_value = xx,
               summary_measure = "xx", #Can choose between mean or median
               shift = xx, #Difference between the null value and the sample mean
               as_extreme_as = xx, #Value of the summary statistic
               direction = "xx", #Specify direction of alternative hypothesis
               number_repetitions = 10000)
```

Notes on the null distribution

Interpretation of the p-value in context of the problem.

Conclusion of the test:

Simulation methods to create a confidence interval

Unlike creation of the null distribution, the bootstrap distribution we use for creating a confidence interval is found by sampling with replacement from the original sample. To create one dot on the bootstrap distribution:

- Write the original values for the variable on n cards; one card for each observational unit.
- Sample with replacement from the cards n times.
- Plot the mean from each resample on the bootstrap distribution.

Use the provided R script file to find a 95% confidence interval.

- Enter the name of the variable for `variable`.
- Enter the appropriate confidence level for `xx`.
- Highlight and run lines 22–25.

```
one_mean_CI(songs$variable, #Enter the name of the variable  
             summary_measure = "mean", #choose the mean or median  
             number_repetitions = 10000, # Number of simulations  
             confidence_level = xx)
```

Notes on the bootstrap distribution

Confidence Interval:

Interpretation of the confidence interval:

6.3.4 Take-home messages

1. We use bootstrapping—sampling with replacement—from the shifted data to generate a null distribution of simulated sample means. In order to ensure that the null distribution is centered at the null value, μ_0 , we shift the data by adding $\mu_0 - \bar{x}$ to each value in the original data set. Note that if this value of the shift is negative, we are shifting the data down; if it is positive, we shift the data up.
2. The mean of the shifted data will equal the null value, μ_0 , but the standard deviation of the shifted data will be the same as the standard deviation of the original data.
3. As in the one proportion scenario, we calculate the p-value for a simulation-based hypothesis test for a single mean by finding the proportion of simulated sample means that are as or more extreme (in the direction of H_A) as the observed sample mean, \bar{x} .

6.3.5 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered.

6.4 Activity 11: Body Temperature

6.4.1 Learning outcomes

- Given a research question involving a quantitative variable, construct the null and alternative hypotheses in words and using appropriate statistical symbols.
- Describe and perform a theory-based hypothesis test for a single mean.
- Interpret and evaluate a p-value for a theory-based hypothesis test for a single mean.

6.4.2 Terminology review

In today's activity, we will analyze quantitative data using theory-based methods. Some terms covered in this activity are:

- Normality
- t -distribution
- Degrees of freedom
- T -score

To review these concepts, see Chapters 11 and 17 in the textbook.

6.4.3 Body Temperature

It has long been reported that the mean body temperature of adults is 98.6°F. There have been a few articles that challenge this assertion. (LUETKEMEIER 2017) In 2018, a sample of 52 Stat 216 undergraduates were asked to report their body temperature. Is there evidence that the average body temperature of Stat 216 undergraduates differs from the known temperature of 98.6°F??

- Observational units:
- Variable:
- Type of variable:

Ask a research question

1. Write out the null hypothesis in proper notation for this study.
2. Write out the alternative hypothesis in words for this study.

In general, the sampling distribution for a sample mean, \bar{x} , based on a sample of size n from a population with a true mean μ and true standard deviation σ can be modeled using a Normal distribution when certain conditions are met.

Conditions for the sampling distribution of \bar{x} to follow an approximate Normal distribution:

- **Independence:** the sample's observations are independent, e.g., are from a simple random sample. (*Remember:* This also must be true to use simulation methods!)
- **Normality Condition:** either the sample observations come from a normally distributed population or we have a large enough sample size. To check this condition, use the following rules of thumb:
 - $n < 30$: If the sample size n is less than 30 and the distribution of the data is approximately normal with no clear outliers in the data, then we typically assume the data come from a nearly normal distribution to satisfy the condition.
 - $30 \leq n < 100$: If the sample size n is between 30 and 100 and there are no particularly extreme outliers in the data, then we typically assume the sampling distribution of \bar{x} is nearly normal, even if the underlying distribution of individual observations is not.
 - $n \geq 100$: If the sample size n is at least 100 (regardless of the presence of skew or outliers), we typically assume the sampling distribution of \bar{x} is nearly normal, even if the underlying distribution of individual observations is not.

Like we saw in Chapter 5, we will not know the values of the parameters and must use the sample data to estimate them. Unlike with proportions, in which we only needed to estimate the population proportion, π , quantitative sample data must be used to estimate both a population mean μ and a population standard deviation σ . This additional uncertainty will require us to use a theoretical distribution that is just a bit wider than the standard Normal distribution. Enter the ***t*-distribution**!

As you can see from Figure 6.1, the *t*-distributions (dashed and dotted lines) are centered at 0 just like a standard Normal distribution (solid line), but are slightly wider. The variability of a *t*-distribution depends on its degrees of freedom, which is calculated from the sample size of a study. (For a single sample of n observations or paired differences, the degrees of freedom is equal to $n - 1$.) Recall from previous classes that larger sample sizes tend to result in narrower sampling distributions. We see that here as well. The larger the sample size, the larger the degrees of freedom, the narrower the *t*-distribution. (In fact, a *t*-distribution with infinite degrees of freedom actually IS the standard Normal distribution!)

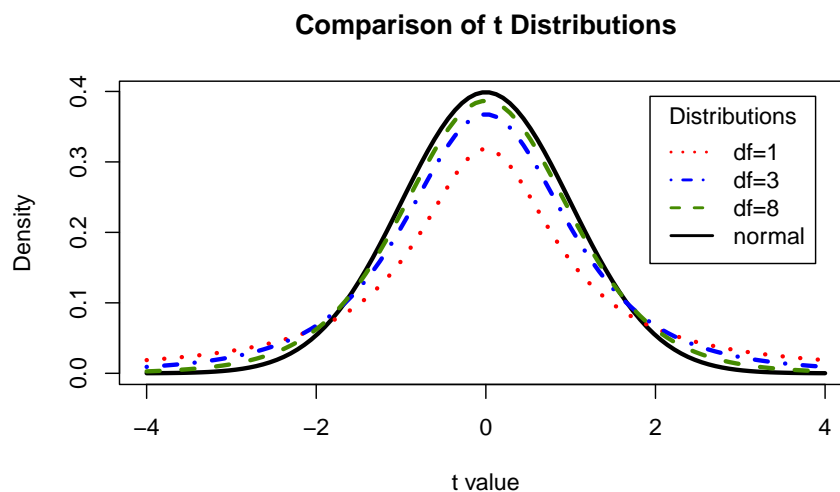


Figure 6.1: Comparison of the standard Normal vs *t*-distribution with various degrees of freedom

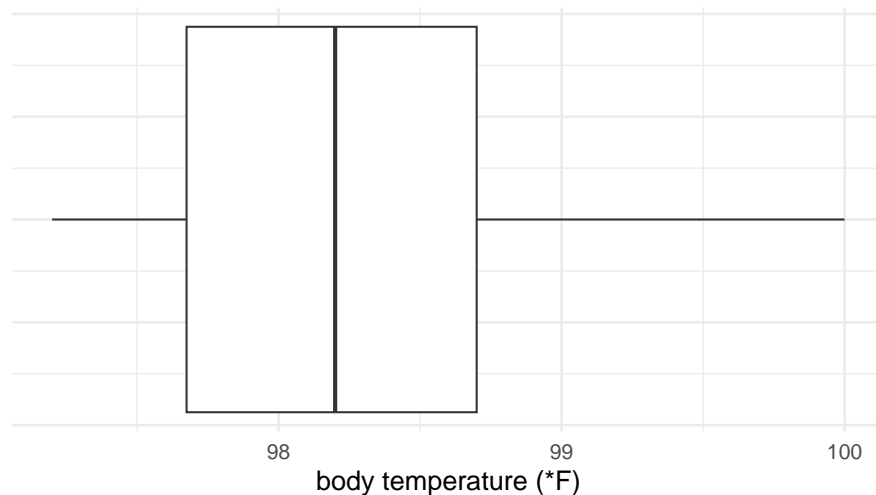
Summarize and visualize the data

The following code is used to create a boxplot of the data.

- Download the R script file and upload to the RStudio server.
- Open the R script file and highlight and run lines 1–14.

```
bodytemp <- read.csv("https://math.montana.edu/courses/s216/data/normal_temperature.csv")
bodytemp %>%
  ggplot(aes(x = Temp))+
  geom_boxplot()+
  labs(title="Boxplot of Body Temperatures for Stat 216 Students",
       x = "body temperature (*F)" +
       theme(axis.text.y = element_blank(),
             axis.ticks.y = element_blank()) # Removes y-axis ticks
```

Boxplot of Body Temperatures for Stat 216 Students



- Highlight and run lines 17 - 18 to get the summary statistics for the variable Temp.

```
bodytemp %>%
  summarise(favstats(Temp))
```

```
#>   min    Q1 median   Q3 max    mean      sd  n missing
#> 1 97.2 97.675  98.2 98.7 100 98.28462 0.6823789 52      0
```

Check theoretical conditions

3. Report the sample size of the study. Give appropriate notation.
4. Report the sample mean of the study. Give appropriate notation.

Verify the independence condition is met:

Verify the normality condition is met to use the theory-based methods:

Use statistical inferential methods to draw inferences from the data

To find the standardized statistic for the mean we will use the following formula:

$$T = \frac{\bar{x} - \mu_0}{SE(\bar{x})},$$

where the standard error of the sample mean is:

$$SE(\bar{x}) = \frac{s}{\sqrt{n}}.$$

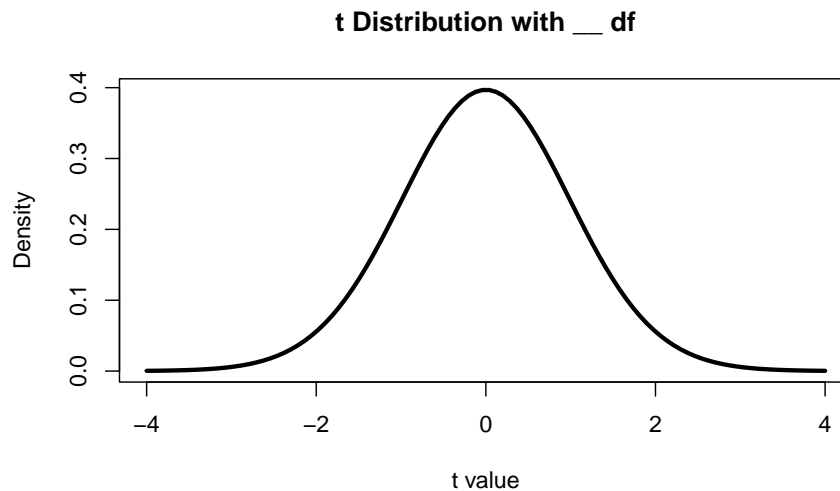
Calculate the standard error of the sample mean.

Interpretation of the standard error in context of the study

Calculate the standardized mean.

We model a single mean with a t -distribution with $n - 1$ degrees of freedom. Calculate the degrees of freedom for this study and use it to fill in the blank in the title of the t -distribution displayed below.

Mark the value of the standardized statistic on the t -distribution and illustrate how the p-value is found.



To find the p-value for the theory-based test in R:

- Enter the value for the standardized statistic for `xx` in the `pt` function.
- Enter the degrees of freedom for `yy` in the `pt` function.
- Highlight and run line 24.

```
2*pt(xx, df=yy, lower.tail=TRUE)
```

5. What does this p-value mean, in the context of the study? Hint: it is the probability of what...assuming what?

6. Write a conclusion to the test in context of the study.

Theory-based methods to create a confidence interval

Next we will calculate a theory-based confidence interval. To calculate a theory-based confidence interval for the a single mean, use the following formula:

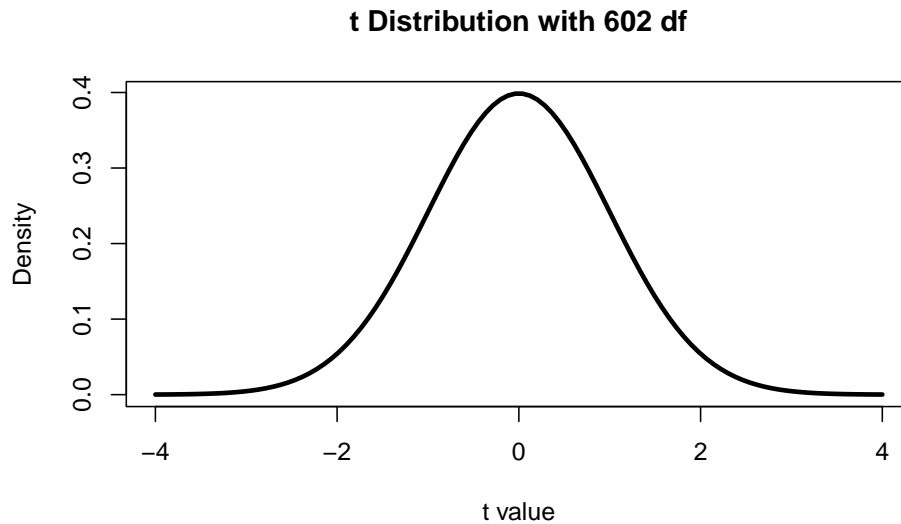
$$\bar{x} \pm t^* \times SE(\bar{x}).$$

We will need to find the t^* multiplier using the function `qt()`.

- Enter the appropriate percentile in the R code to find the multiplier for a 95% confidence interval.
- Enter the degrees of freedom for `yy`. *The degrees of freedom for a single mean is $n - 1$.*
- Highlight and run line 31.

```
qt(percentile, df = yy, lower.tail=TRUE)
```

Mark on the t -distribution found below the values of $\pm t^*$. Draw a line at each multiplier and write the percentiles used to find each.



Calculate the margin of error using theory-based methods

Calculate the confidence interval for the true mean using theory-based methods.

7. Interpret the confidence interval in context of the study.
8. Can we generalize the results of the study to all adults? Explain your answer.

6.4.4 Take-home messages

1. In order to use theory-based methods for a quantitative variable, the independent observational units and normality conditions must be met.
2. In order to find a theory-based p-value, we use R to calculate the area under a t -distribution with $n - 1$ degrees of freedom (df) that is at or more extreme than the observed T -score. To find a two-sided p-value using theory-based methods we need to multiply the one-sided p-value by 2.
3. A t^* multiplier is found by obtaining the bounds of the middle X% (X being the desired confidence level)

of a t -distribution with $n - 1$ df.

6.4.5 Additional notes

Use this space to summarize your thoughts and take additional notes on today's activity and material covered

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