

Inference Review

Hypothesis Testing

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01.15.20

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Announcements

- Complete [surveys and consent form](#) by Wed at 11:59p
- [Reading for next Wednesday](#).
- Labs start tomorrow!
- No class Monday - Martin Luther King, Jr. Holiday
- Find more info about statistics related events on [Sakai](#)

Today's Agenda

- Calculating & interpreting hypothesis tests
- Drawing conclusions using hypothesis tests and confidence intervals

Sesame Street

- *Sesame Street* is a television series designed to teach children ages 3-5 basic education skills such as reading (e.g. the alphabet) and math (e.g. counting)
- Today we are going to analyze data from an [study conducted by the Educational Testing Service](#) in the early 1970s to test the effectiveness of the program.



Sesame Street study

- Children from 6 locations around the United States (including Durham!) participated in the 26-week study. The children were split into two groups (treatment):
 - **Group 1:** Those who were encouraged to watch the show (assume watched regularly)
 - **Group 2:** Those who didn't get encouragement to watch the show (assume didn't watch regularly)
- Each child was given a test before and after the study to measure their knowledge of basic math, reading, etc.
- We will focus on the change in reading (identifying letters) scores (change)



[Sesame Street Data - Full Description](#) Original Study: *Ann Bogatz, Gerry & Ball, Samuel. (1971). The Second Year of Sesame Street: A Continuing Evaluation. Volume 1. vols. 1 & 2.*

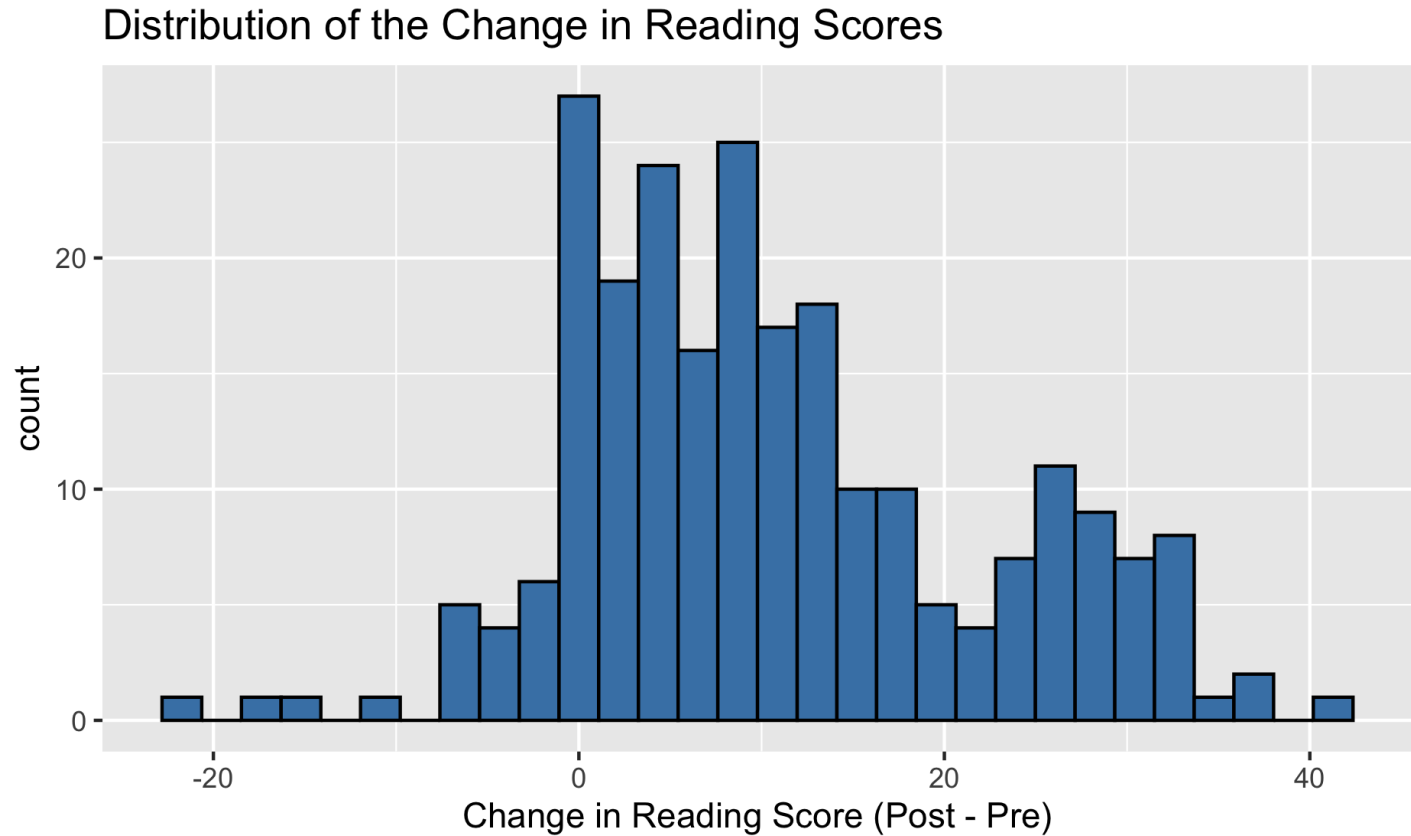
Let's look at the data

sesame_street.csv is available in the datasets repo on GitHub.

```
sesame_street %>%  
  slice(1:10)
```

```
## # A tibble: 10 x 4  
##   treatment      prelet postlet change  
##   <chr>         <dbl>   <dbl>   <dbl>  
## 1 Encouraged      23      30      7  
## 2 Encouraged      26      37     11  
## 3 Not Encouraged  14      46     32  
## 4 Not Encouraged  11      14      3  
## 5 Not Encouraged  47      63     16  
## 6 Not Encouraged  26      36     10  
## 7 Not Encouraged  12      45     33  
## 8 Encouraged      48      47     -1  
## 9 Encouraged      44      50      6  
## 10 Encouraged     38      52     14
```

Exploratory Data Analysis - Univariate



Exploratory Data Analysis - Univariate

- Calculate summary statistics for change

```
sesame_street %>%  
  summarise(n = n(), min = min(change), median = median(change), n  
            IQR = IQR(change),  
            mean = mean(change), std_dev = sd(change))
```

```
## # A tibble: 1 x 7  
##       n    min median    max   IQR  mean std_dev  
##   <int> <dbl>  <dbl> <dbl> <dbl> <dbl>  <dbl>  
## 1    240   -22      9    41    15  10.8   11.2
```

95% CI for mean change in reading score

The 95% confidence interval for the mean change in reading score is

[9.384, 12.224]

- Interpret the interval at <http://bit.ly/sta210-sp20-CI-2>
- Use **NetId@duke.edu** for your email address.
- You are welcome (and encouraged!) to discuss these questions with 1 - 2 people around you, but **each person** must submit a response.

03 : 00

Confidence Interval Recap

Hypothesis Tests

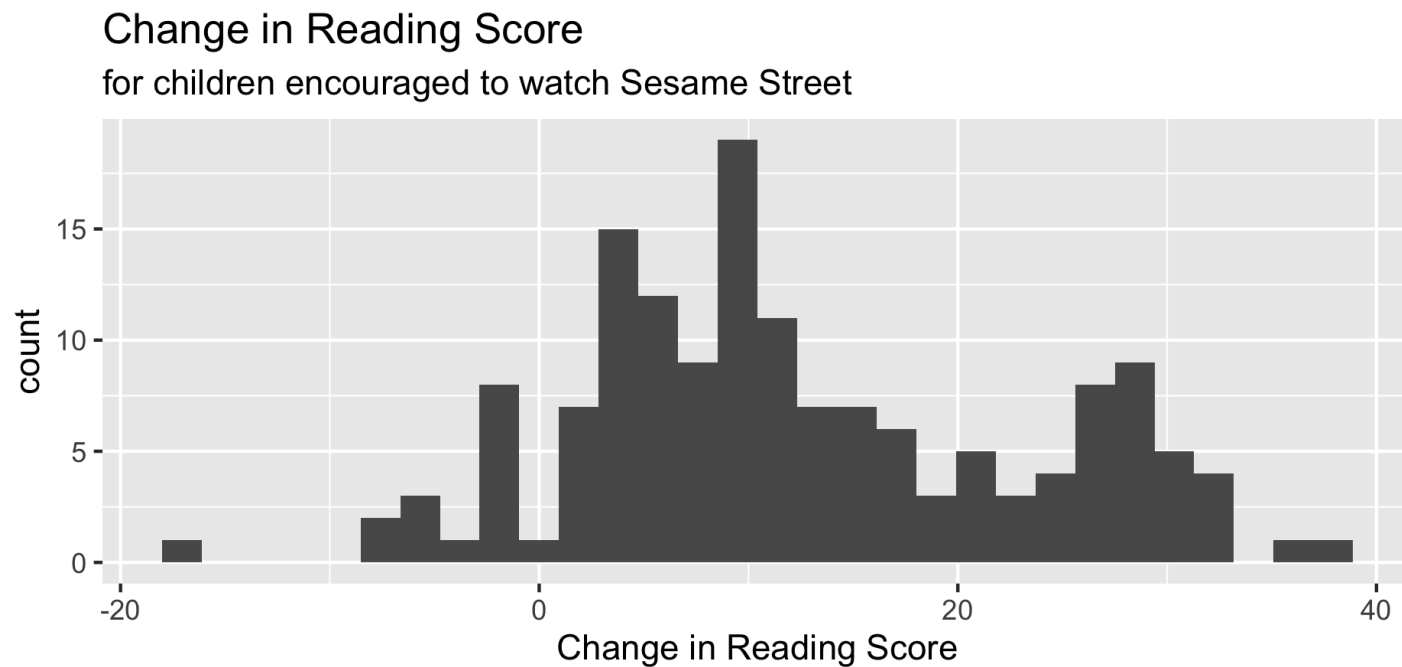
Question we want to answer

- Let's focus on the children who were encouraged to watch *Sesame Street*
- In general, those children watched the show regularly, so let's see if the show impacted their reading skills

The mean change in reading scores after 26 weeks for children ages 3 - 5 is 11.

Is there evidence that mean change in reading score for children encouraged to watch *Sesame Street* is "significantly" different from the mean change in reading score for all children?

Let's look at the data



```
## # A tibble: 1 x 3
##       n mean std_dev
##   <int> <dbl>   <dbl>
## 1   152  12.5    10.7
```

Outline of a Hypothesis Test

- Identify the parameter of interest.
- Identify a null hypothesis, H_0 , that represents the baseline
- Set an alternative hypothesis, H_a , that represents the research question, i.e. what you're testing
- Conduct a hypothesis test under the assumption that the null hypothesis is true and calculate a p-value
 - The **p-value** is the probability of getting the observed outcome or a more extreme outcome given the null hypothesis is true

Outline of a Hypothesis Test

- Assess the p-value. A small p-value means...
 - a. The assumed (null) hypothesis is incorrect
 - b. The assumed (null) hypothesis is correct and a rare event has occurred
- State a conclusion about the hypothesis based on the assessment of the p-value
 - Since event (b) is by definition rare, we will conclude a "small" p-value indicates that there is sufficient evidence to claim that the assumed hypothesis is false.
 - In other words, the data are not consistent with the assumed hypothesis
 - When the p-value is "not small", we will conclude that there is not sufficient evidence to claim the assumed hypothesis is false.

Identify parameter & hypotheses

- **Null hypothesis, H_0** : This is the baseline hypothesis, i.e. the "there is nothing going on" hypothesis.
 - The mean change in reading score for children encouraged to watch the show is 11 (same as the mean for all children)
- **Alternative hypothesis, H_a** : This is typically what you want to show, i.e. the "there is something going on" hypothesis
 - The mean change in reading score for children encouraged to watch the show not 11 (different from the mean for all children)

$$H_0 : \mu = 11$$

$$H_a : \mu \neq 11$$

Distribution \bar{x} under H_0

- We want to draw conclusions about μ , so we'll use our best guess \bar{x}
- Recall from the Central Limit Theorem, when certain conditions are met (they are!), we know that

$$\bar{x} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$$

- We conduct a hypothesis test under the assumption that H_0 is true, so for this test

$$\bar{x} \sim N\left(11, \frac{\sigma}{\sqrt{n}}\right)$$

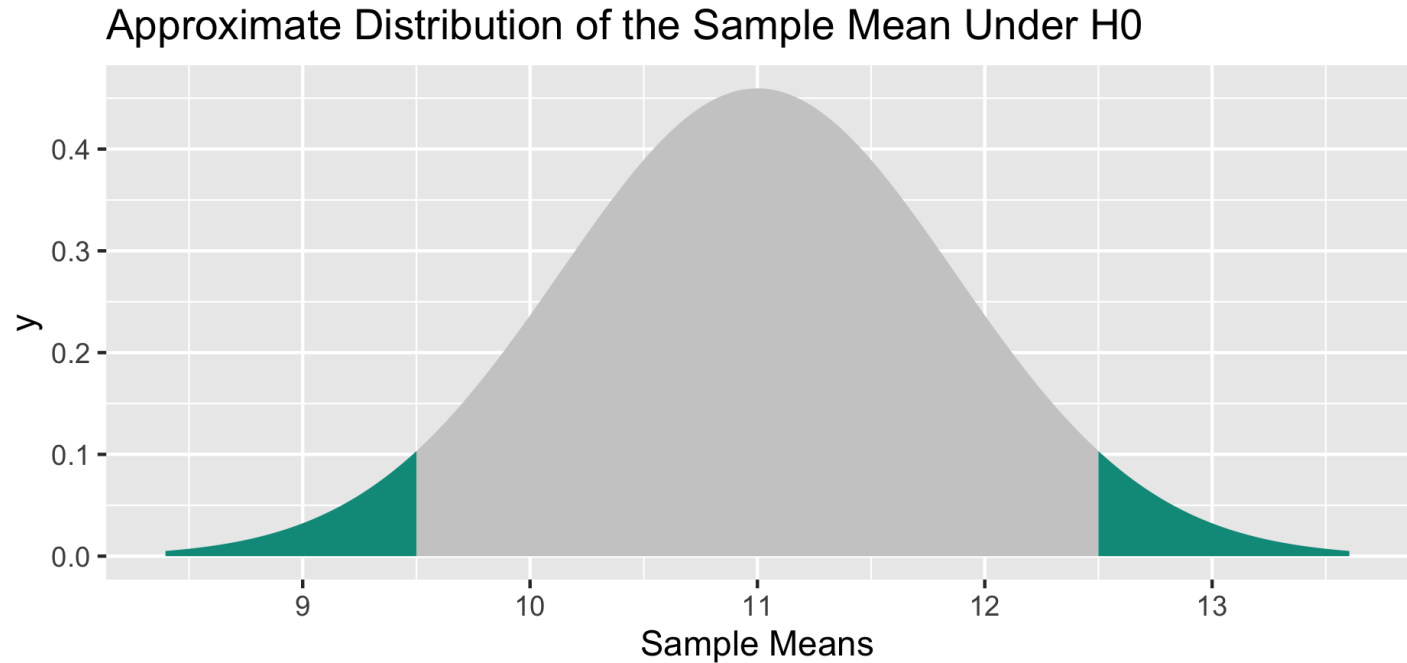
Distribution \bar{x} under H_0

- We don't know σ , so we can use the standard error s/\sqrt{n} to approximate σ/\sqrt{n} .
- Thus, putting it all together, we know

$$\bar{x} \approx N\left(11, \frac{10.7}{\sqrt{152}}\right)$$

Given $\bar{x} \approx N\left(11, \frac{10.7}{\sqrt{152}}\right)$, what is the probability of observing a mean change in score at least 1.5 points away from the center (11) in a random sample of 152 children ages 3 - 5?

Visualize



The shaded area represents the (approximate) probability of obtaining a sample mean at least as far away from the center as the one we observed given the true mean change is 11.

Test Statistic

- Let's quantify how "unusual" our observed sample mean is given $H_0 : \mu = 11$ is true
- We'll begin by calculating how "far away" the observed mean is from the center of the distribution under H_0
- The **test statistic** is the number of standard errors the observed value is from the hypothesized value. The general form of the test statistic is

$$\frac{\text{observed value} - \text{hypothesized value}}{SE}$$

$$t = \frac{\bar{x} - \mu_0}{\frac{s}{\sqrt{n}}} = \frac{12.5 - 11}{\frac{10.7}{\sqrt{152}}} \approx 1.728$$

where the test statistic follows the t distribution with $n - 1$ df

Motivating the p-value

- We got a test statistic of 1.728. In other words...

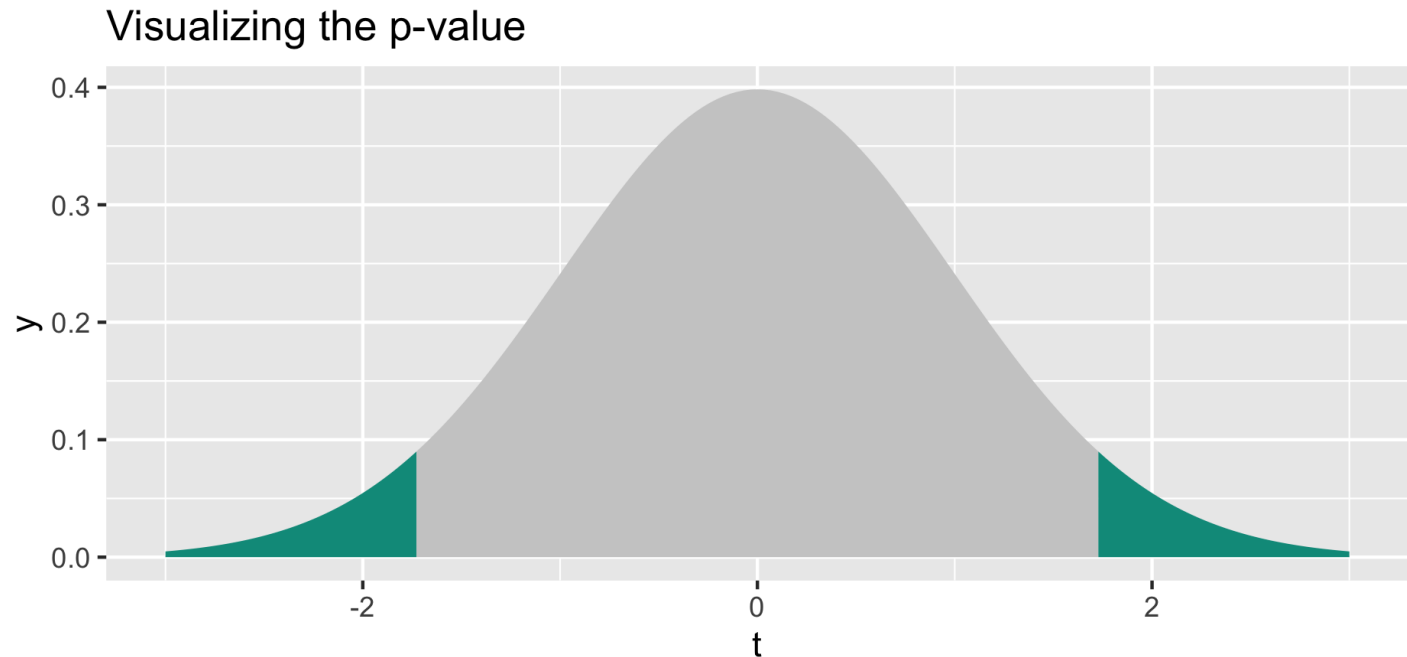
Given $\bar{x} \approx N\left(11, \frac{10.7}{\sqrt{152}}\right)$, what is the probability of observing a mean change in score at least 1.5 points away from the center (11) in a random sample of 152 children ages 3 - 5?



Given the t distribution with 151 degrees of freedom, what is the probability of observing a test statistic with magnitude 1.728 or larger?

p-value

Given the t distribution with 151 degrees of freedom, what is the probability of observing a test statistic with magnitude 1.728 or larger?



```
(p_value <- 2 * pt(-1.728, 151))
```

```
## [1] 0.08603259
```

General guide for interpreting the p-value

Magnitude of p-value	Interpretation
p-value < 0.01	strong evidence against H_0
0.01 < p-value < 0.05	moderate evidence against H_0
0.05 < p-value < 0.1	weak evidence against H_0
p-value > 0.1	effectively no evidence against H_0

Note: These are general guidelines. The strength of evidence depends on the context of the problem.

Drawing the conclusion: Part 1

- A threshold can be used to decide whether or not to reject H_0 in favor of the alternative H_a
- This threshold is called the **significance level** and is usually denoted by α
- If the p-value is less than α , then we conclude there is sufficient evidence against H_0 — and we **reject the null hypothesis**
- Otherwise, we conclude that there isn't sufficient evidence against H_0 and **fail to reject the null hypothesis**

Don't just rely on p-values

- Do not rely strictly on the p-value and significance level to make a conclusion!
- Suppose the significance level is 0.05
 - If the p-value is 0.05001, we fail to reject H_0
 - If the p-value is 0.04999, we reject H_0
- 0.05001 and 0.04999 are practically the same, yet they led to different conclusions.

Use confidence intervals and other statistical summaries to provide more context about the results.

t-test for *Sesame Street* data

```
enc <- sesame_street %>%  
  filter(treatment == "Encouraged")  
  
t.test(enc$change, mu = 11, conf.level = 0.9,  
       direction = "two.sided")
```

```
##  
##      One Sample t-test  
##  
## data:  enc$change  
## t = 1.7226, df = 151, p-value = 0.08701  
## alternative hypothesis: true mean is not equal to 11  
## 90 percent confidence interval:  
##  11.05883 13.94117  
## sample estimates:  
## mean of x  
##      12.5
```

In-class exercise

- Answer the questions: <http://bit.ly/sta210-sp20-ht>
- Use **NetId@duke.edu** for your email address.
- You are welcome (and encouraged!) to discuss these questions with 1 - 2 people around you, but **each person** must submit a response.

04:00

Conclusion

p-value: 0.087

90% confidence interval: [11.059, 13.941]

- Using a significance level of 0.1, what is your conclusion from the test?
- Suppose you are advising a group of educators about whether they should spend additional time and money to encourage children to watch *Sesame Street*. Based on these results, would you advise the educators to spend the resources? Why or why not?

Inference for difference in means $\mu_1 - \mu_2$

By the Central Limit Throem, when the conditions are met,

$$(\bar{x}_1 - \bar{x}_2) \sim N\left(\mu_1 - \mu_2, \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}\right)$$

- We don't know σ_1 and σ_2 in practice, so we use the **standard error** in all calculations.

$$SE(\bar{x}_1 - \bar{x}_2) = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

Inference for difference in means $\mu_1 - \mu_2$

Confidence Interval to estimate $\mu_1 - \mu_2$

$$(\bar{x}_1 - \bar{x}_2) \pm t_{df}^* \times \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

t^* follows a t distribution with degrees of freedom computed in R.

Inference for difference in means $\mu_1 - \mu_2$

Hypothesis Test: Is there a difference in the means between Group 1 and Group 2?

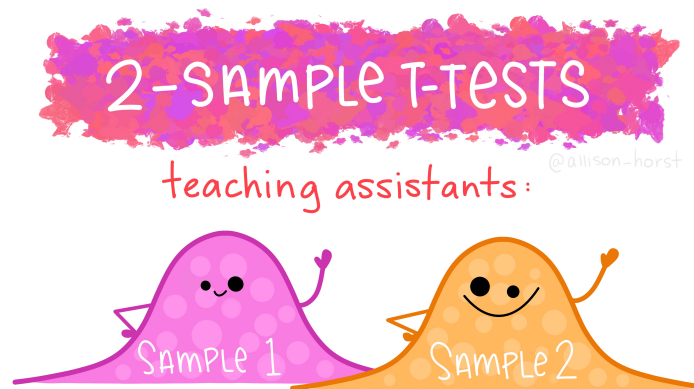
- Null hypothesis: $H_0 : \mu_1 - \mu_2 = 0$
- Test statistic:

$$\frac{(\bar{x}_1 - \bar{x}_2) - 0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

- p-value: Calculated using the t distribution with degrees of freedom computed in R.

Additional Resources

- Discussion in the scientific community about p-values: "[Scientists rise up against statistical significance](#)" in *Nature*
- Fun review of two-sample tests by [@allison_horst](#):
https://twitter.com/allison_horst/status/1216411185240690688



By Thursday at noon

- Make sure you are a member of the [course organization on GitHub](#)
- Make sure you have access to RStudio
- If you are using RStudio on your local machine, make sure you have git configured and you can knit a PDF (need a Latex editor installed)