

# More on Permutation tests

Stat 250

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# Questions?

What questions do you have from last class,  
including the extra examples?

# Implementation

# Data overview

Data frame with 47 rows and two columns

```
dplyr::glimpse(motivation)
```

```
Rows: 47
```

```
Columns: 2
```

```
$ Score      <dbl> 5.0, 5.4, 6.1, 10.9, 11.8, 12.0, 12.3,  
14.8, 15.0, 16.8, 17...
```

```
$ Treatment <fct> Extrinsic, Extrinsic, Extrinsic,  
Extrinsic, Extrinsic, Extri...
```

# 1. Calculate the observed difference in means

```
group_means <- mosaic::mean(~Score | Treatment, data = motivation)
group_means
```

```
Extrinsic Intrinsic
15.73913  19.88333
```

```
observed <- group_means["Intrinsic"] - group_means["Extrinsic"]
observed
```

```
Intrinsic
4.144203
```

## 2. Use a for loop to run the permutations

```
1 y <- motivation$Score # vector with response variable
2 n <- nrow(motivation) # total sample size
3 ngrp1 <- 24           # sample size for group 1
4 N <- 10^4 - 1         # number of resamples
5 result <- numeric(N)  # place to store results
6
7 for(i in 1:N) {
8   index <- sample(n, size = ngrp1, replace = FALSE)
9   result[i] <- mean(y[index]) - mean(y[-index])
10 }
```

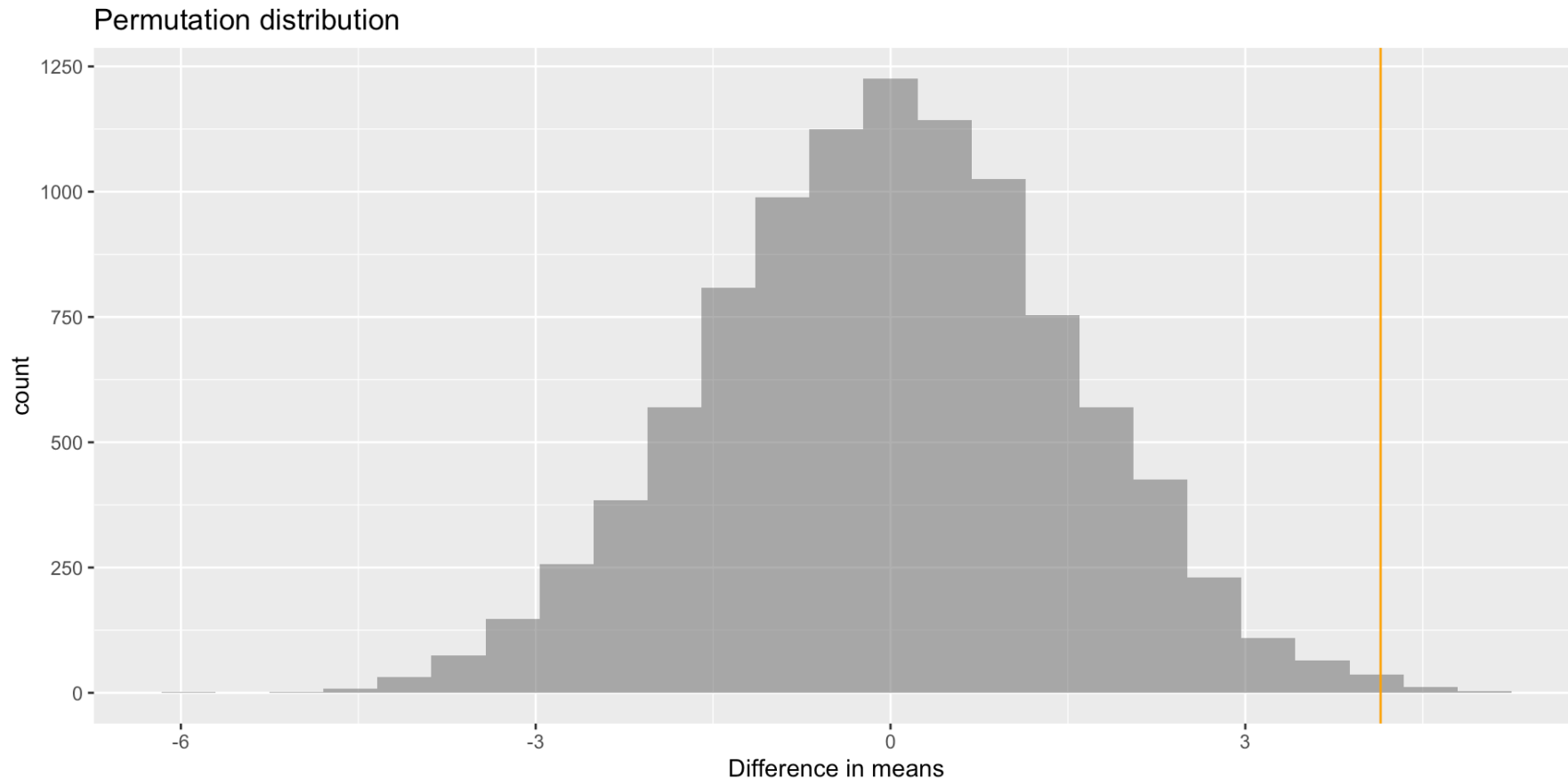
`result` is a vector

```
head(result)
```

```
[1] -2.2331522 -3.5614130 -2.7780797 -1.6626812
0.1849638 -2.6503623
```

### 3. Plot the permutation (null) distribution

```
gf_histogram(~result, title = "Permutation distribution"  
             xlab = "Difference in means") |>  
gf_vline(xintercept = ~observed, color = "orange")
```



## 4. Calculate the p-value

⚠ Be sure to **look at the alternative hypothesis** to select the inequality (tail)

```
(sum(result >= observed) + 1) / (N + 1)
```

```
[1] 0.0025
```

To get a two-sided p-value, multiple by 2

```
2 * (sum(result >= observed) + 1) / (N + 1)
```

```
[1] 0.005
```



# Permutation test for difference in proportions

Example: Hearing loss in U.S. teens, 1988-1994  
vs. 2005-06

```
glimpse(hearing_loss)
```

```
Rows: 4,699
```

```
Columns: 2
```

```
$ year   <chr> "2005-2006", "2005-2006", "2005-2006",  
"2005-2006", "2005-2006", ...  
$ hloss  <chr> "yes", "yes", "yes", "yes", "yes", "yes",  
"yes", "yes", "yes", ...
```

# 1. Calculate the observed difference in means

```
group_props <- mosaic::prop(~hloss | year, data = hearing_loss, success = "yes")
group_props
```

```
prop_yes.1988-1994 prop_yes.2005-2006
0.1639344         0.1880294
```

```
observed <- group_props["prop_yes.1988-1994"] - group_props["prop_yes.2005-2006"]
observed
```

```
prop_yes.1988-1994
-0.02409494
```

## 2. Use a for loop to run the permutations

```
1 y <- hearing_loss$hloss      # vector with response variable
2 n <- nrow(hearing_loss)      # total sample size
3 ngrp1 <- 2928                 # sample size for group 1
4 N <- 10^4 - 1                # number of resamples
5 result <- numeric(N)         # place to store results
6
7 for(i in 1:N) {
8   index <- sample(n, size = ngrp1, replace = FALSE)
9   result[i] <- mean(y[index] == "yes") - mean(y[-index] ==
10 }
```

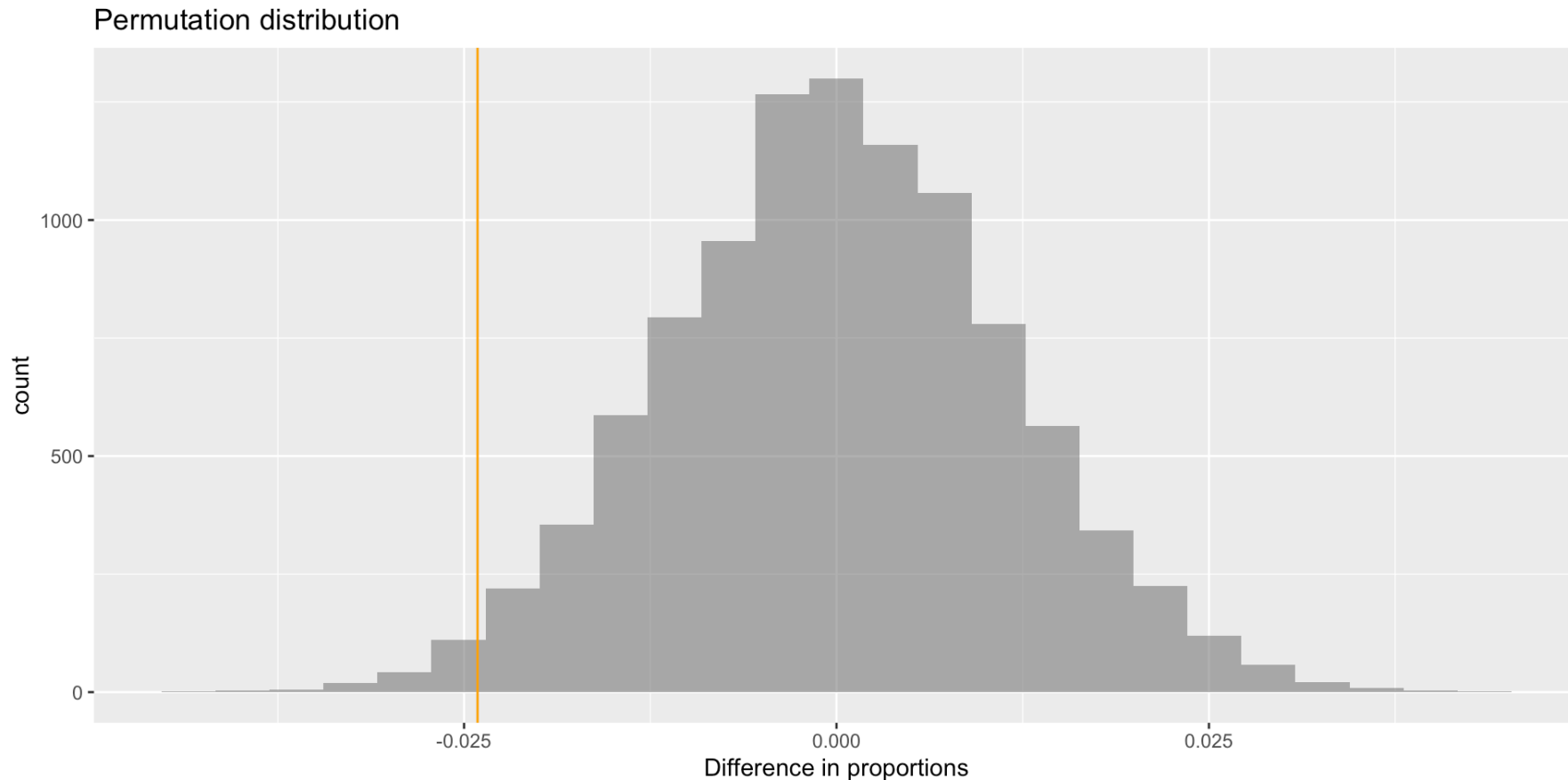
`result` is a vector

```
head(result)
```

```
[1] 0.004902914 -0.013220742 -0.023188753 0.007621462
0.013964742
[6] -0.025001119
```

### 3. Plot the permutation (null) distribution

```
gf_histogram(~result, title = "Permutation distribution"  
  xlab = "Difference in proportions") |>  
  gf_vline(xintercept = ~observed, color = "orange")
```



## 4. Calculate the p-value

⚠ Be sure to **look at the alternative hypothesis** to select the inequality (tail)

```
(sum(result <= observed) + 1) / (N + 1)
```

```
[1] 0.0184
```

To get a two-sided p-value, multiple by 2

```
2 * (sum(result <= observed) + 1) / (N + 1)
```

```
[1] 0.0368
```

# Reproducibility

Every time you run sampling code you will received a different random sample **unless** you set the seed (random number generator state).

Add `set.seed()` to the code chunk where you sample (or at the top of an .Rmd file)

```
set.seed(1234) # Choose some large integer
```

# Your turn

Work through the R examples.

# Matched pairs

Data are **paired** if the groups being compared are clearly *linked*

- Two measurements on each case
- Twin studies
- Each case is matched with a similar case, and one case in each pair is given each treatment
- Other situations where the data are naturally paired



# Your turn

Which of the following scenarios should be analyzed as paired data?

1. Students take an MCAT prep course. Their before and after scores are compared.
2. 20 first-year and 20 second-year students in class take a midterm. We compare their scores.
3. A group of freshman are asked about the quality of food on campus. A year later, the same students are asked this question again. Do student's opinions change over time?

# Is it safe to look at social media while driving?

- Previous research on smart phone use while driving has primarily focused on phone calls and texting.
- Study looked at the effects of different smart phone tasks on car-following performance in a driving simulator.
- Drivers performed driving only baseline simulation
- Drivers performed other phone tasks: texting, reading Facebook posts, exchanging photos on Snapchat, viewing updates on Instagram
- Brake reaction times (in seconds) recorded

# Your turn

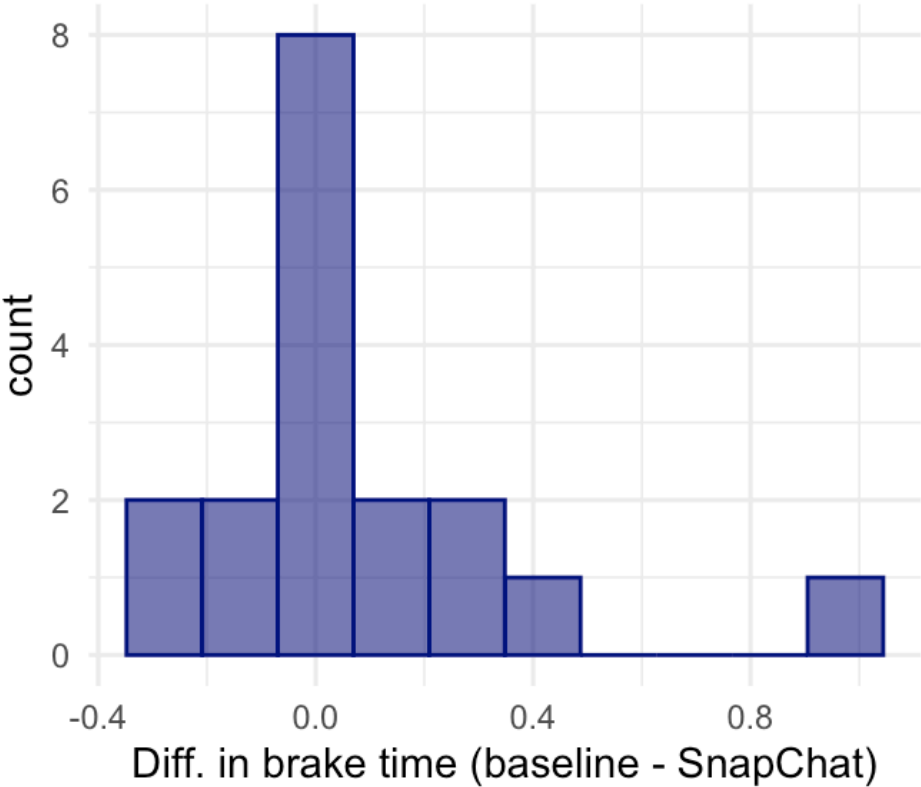
- Is it safe to look at social media while driving?
- What are logical null and alternative hypotheses for this research question?

# Test statistic

Subject	Baseline	SnapChat	Diff
1	0.863	0.865	<b>0.002</b>
2	0.847	0.783	<b>-0.063</b>
3	0.836	0.808	<b>-0.028</b>
4	0.655	1.010	<b>0.354</b>
5	0.900	0.837	<b>-0.063</b>
6	0.957	1.175	<b>0.218</b>
7	0.780	0.817	<b>0.037</b>
8	0.954	0.861	<b>-0.094</b>
9	0.970	0.717	<b>-0.253</b>
10	1.102	1.141	<b>0.039</b>
11	0.925	0.583	<b>-0.342</b>
12	0.833	0.883	<b>0.050</b>
13	0.833	0.995	<b>0.161</b>
14	0.773	0.837	<b>0.064</b>
15	0.914	1.008	<b>0.095</b>
16	0.858	1.137	<b>0.278</b>
17	0.822	1.733	<b>0.911</b>
18	0.963	0.883	<b>-0.079</b>

- For a matched pairs experiment, look at the differences between responses for each unit (pair)
- Compute a new variable for differences
- We use the mean difference as our test statistic

# EDA



mean	sd	n
0.072	0.27	18

min	Q1	median
-0.342	-0.063	0.038

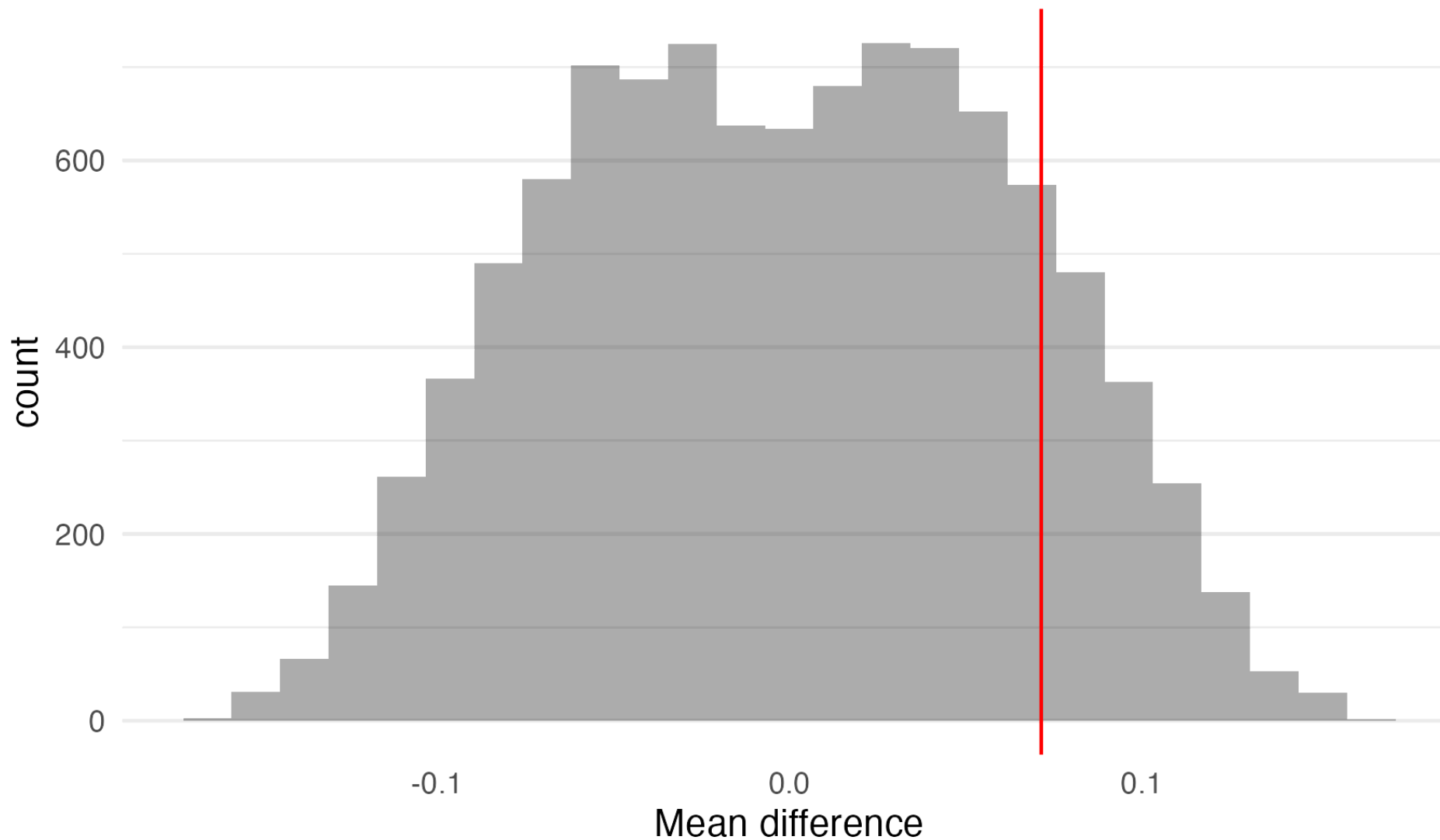
# Generating the null distribution

Subject	Baseline	SnapChat	Diff
1	0.863	0.865	<b>0.002</b>
2	0.847	0.783	<b>-0.063</b>
3	0.836	0.808	<b>-0.028</b>
4	0.655	1.010	<b>0.354</b>
5	0.900	0.837	<b>-0.063</b>
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18	0.963	0.883	<b>-0.079</b>

- Can't permute one of the columns and recalculate the difference
- Instead, **randomly select a sign (+/-) for each difference**
- Use the mean difference as the test statistic

# Permutation distribution

Assuming no difference, based on 9,999 trials



# Strength of evidence

- 1481 simulated test statistics exceed the observed
- 9999 total statistics are in the null distribution
- What's the  $p$ -value?
- Is there a statistical discernible difference in braking time?



# Implementation

Data frame with 18 rows and 6 columns

```
glimpse(brake)
```

```
Rows: 18
Columns: 6
$ Subject    <dbl> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13,
14, 15, 16, 17, 18
$ Baseline   <dbl> 0.863358, 0.846684, 0.836120, 0.655340,
0.900022, 0.956534, ...
$ Facebook   <dbl> 1.254010, 1.100017, 1.020859, 0.864210,
0.793354, 1.252108, ...
$ Texting    <dbl> 1.011476, 0.900027, 1.063912, 0.973921,
0.856002, 1.178000, ...
$ Instagram  <dbl> 0.962927, 0.600018, 0.946600, 0.725775,
0.816683, 1.134345, ...
$ SnapChat   <dbl> 0.865000, 0.783353, 0.808357, 1.009547,
0.836783, 1.175025, ...
```

Add a **Diff** column

```
brake <- mutate(brake, Diff = Baseline - SnapChat)
```

# Implementation

```
1 Diff <- brake$Diff      # vector of differences
2 observed <- mean(Diff)   # observed test stat
3 n <- length(Diff)        # sample size
4 N <- 10^4 - 1            # no. of permutation resamples
5
6 set.seed(120)            # set seed for reproducibility
7 result <- numeric(N)     # place to store the results
8
9 for (i in 1:N){
10     swap <- sample(c(-1,1), n, replace = TRUE)
11     result[i] <- mean(swap * Diff)
12 }
```