R Coding Demonstration Week 11: Hypothesis Tests

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Introduction

- Today we're going to cover some tools for exploring bivariate relationships.
- We'll use the data from the Broockman & Kalla (2016) transphobia study.
- · Basic summary of experiment:
 - · Randomly assigned door-to-door canvassers to two conditions
 - Conditions: perspective-taking script (treatment) or recycling script (placebo)
 - Follow up surveys at 3 days, 3 weeks, 6 weeks, and 3 months.

Data

phobia <- read.csv("data/transphobia_all.csv")</pre>

Variable Name	Description
age	Age of the respondent in years
female	1=respondent marked "Female" on voter registration, 0 otherwise
voted_gen_14	1 if respondent voted in the 2014 general election
voted_gen_12	1 if respondent voted in the 2012 general election
treat_ind	1 if respondent was assigned to treatment, 0 for control
racename	character name of racial identity indicated on voter file
democrat	1 if respondent is a registered Democrat
therm_trans_t0	0-100 feeling therm. about transgender people at baseline
therm_trans_tX	0-100 feeling therm. about transgender people in Wave X after treatment
therm_obama_t0	0-100 feeling therm. about Barack Obama at baseline
therm_obama_tX	0-100 feeling therm. about Barack Obama in Wave X after treatment

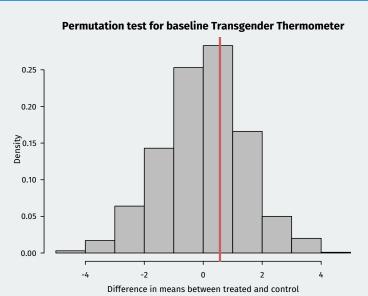
Permutation tests

- A common way to conduct inference in an experiment is called a permutation test.
 - Null hypothesis of no treatment effect: $H_0: Y_i(1) = Y_i(0)$
 - · Doesn't matter if we flip treatment for someone.
 - But different allocation of treatment will lead to different estimated ATEs by random chance.
- · Permutation test:
 - 1. Calculate observed difference in means.
 - 2. Randomly shuffle treatment indicator across units.
 - 3. Calculate difference in means of shuffled data.
 - 4. Repeat 2-3 many times and compare observed effect to this distribution.
- Shuffled distribution = reference distribution.
 - p-value: what proportion of effects would we see if treatment had no effect?

Conduct a permutation test for the "effect" of treatment on baseline thermometer scores for transgender people. Show a histogram of the reference distribution and plot a line that indicates where the observed difference in means is.

```
n sims <- 1000
n_obs <- nrow(phobia)</pre>
trans t0 test <- rep(NA, times = n sims)
trans t0 est <- mean(phobia$therm trans t0[phobia$treat ind == 1]) -
 mean(phobia$therm trans t0[phobia$treat ind == 0])
for (i in 1:n sims) {
  shuff treat <- sample(phobia$treat ind, size = n obs,
                        replace = TRUE)
  trans t0 test[i] <- mean(phobia$therm trans t0[shuff treat == 1]) -
    mean(phobia$therm trans t0[shuff treat == 0])
hist(trans t0 test, col = "grey", freq = FALSE,
     main = "Permutation test for baseline Transgender Thermometer",
     xlab = "Difference in means between treated and control")
abline(v = trans t0 est, lwd = 3, col = "indianred")
```

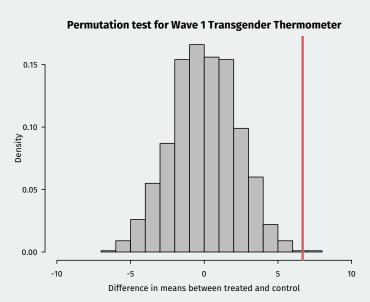
Answer 1 (plot)



Conduct a permutation test for the effect of treatment on thermometer scores for transgender people in wave 1 (therm_trans_t1). Note that there are missing values in wave 1, so be sure to drop those from your analyses.

```
trans t1 test <- rep(NA, times = n sims)</pre>
trans t1 est <- mean(phobia$therm_trans_t1[phobia$treat_ind == 1], na.rm = TRUE),
 mean(phobia$therm trans t1[phobia$treat ind == 0], na.rm = TRUE)
for (i in 1:n sims) {
 shuff treat <- sample(phobia$treat ind. size = n obs.</pre>
                        replace = TRUE)
 trans t1 test[i] <- mean(phobia$therm trans t1[shuff treat == 1], na.rm = TRUE
    mean(phobia$therm trans t1[shuff treat == 0], na.rm= TRUE)
hist(trans t1 test, col = "grey", freq = FALSE, xlim = c(-10, 10),
    main = "Permutation test for Wave 1 Transgender Thermometer".
    xlab = "Difference in means between treated and control")
abline(v = trans_t1_est, lwd = 3, col = "indianred")
```

Answer 2 (plot)



For the baseline thermometer score permutation test, calculate the one-sided and two-sided p-values for the estimated effect.

```
## one-sided
mean(trans_t0_test > trans_t0_est)

## [1] 0.358

## two-sided
mean(trans_t0_test > trans_t0_est) + mean(trans_t0_test < -trans_t0_est)

## [1] 0.679</pre>
```

For the wave 1 thermometer score permutation test, calculate the one-sided and two-sided p-values for the estimated effect.

```
## one-sided
mean(trans_t1_test > trans_t1_est)

## [1] 0.001

## two-sided
mean(trans_t1_test > trans_t1_est) + mean(trans_t1_test < -trans_t1_est)

## [1] 0.001</pre>
```

Conduct a t-test of the difference in average baseline thermometer scores between the treated and control group with the following null hypothesis:

$$\cdot H_0: \mu_T - \mu_C = 0$$

where μ_T is the population mean of the potential outcomes under treatment. What are some differences between this null and the null hypothesis from the permutation test?

```
t.test(phobia$therm_trans_t0[phobia$treat_ind == 1],
    phobia$therm_trans_t0[phobia$treat_ind == 0])
```

```
##
## Welch Two Sample t-test
##
## data: phobia$therm_trans_t0[phobia$treat_ind == 1] and phobia$therm_trans_
## t = 0.4, df = 1820, p-value = 0.7
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -2.12 3.26
## sample estimates:
## mean of x mean of y
## 53.6 53.0
```

Conduct a t-test of the difference in average wave 1 thermometer scores between the treated and control group.

```
t.test(phobia$therm_trans_t1[phobia$treat_ind == 1],
    phobia$therm_trans_t1[phobia$treat_ind == 0])
```

```
##
## Welch Two Sample t-test
##
## data: phobia$therm_trans_t1[phobia$treat_ind == 1] and phobia$therm_trans_
## t = 3, df = 565, p-value = 0.004
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 2.11 11.24
## sample estimates:
## mean of x mean of y
## 60.7 54.1
```