Lab 5

Stat C131A

September 27 and 29, 2021

Welcome to Lab 5! In this lab, we will:

- Learn about how to calculate confidence intervals (CIs) for t-tests.
- Learn how to create confidence interval using bootstrap methods.
- Learn basic syntax of linear regression.

In this lab, we will look at a comic book data created by FiveThirtyEight for their story *Comic Books Are Still Made By Men, For Men And About Men*, where they claimed that Comic books vastly under-represent women. Specifically, they said that characters who are women have lower appearance counts after analyzing the Marvel and DC dataset. Well, is that true? Let's apply our testing methods to their analysis!

```
# Read in data.
marvel <- read.csv("marvel-wikia-data.csv")</pre>
```

The data marvel-wikia-data.csv comes from Marvel Wikia. It has the following variables:

- page_id: The unique identifier for that characters page within the wikia
- name: The name of the character
- urlslug: The unique url within the wikia that takes you to the character
- ID: The identity status of the character (Secret Identity, Public identity, [on marvel only: No Dual Identity])
- ALIGN: If the character is Good, Bad or Neutral
- EYE: Eye color of the character
- HAIR: Hair color of the character
- SEX: Sex of the character (e.g. Male, Female, etc.)
- GSM: If the character is a gender or sexual minority (e.g. Homosexual characters, bisexual characters)
- ALIVE: If the character is alive or deceased
- APPEARANCES: The number of appearances of the character in comic books (as of Sep. 2, 2014. Number will become increasingly out of date as time goes on.)
- FIRST APPEARANCE: The month and year of the character's first appearance in a comic book, if available
- YEAR: The year of the character's first appearance in a comic book, if available

We will focus on two columns SEX and APPEARANCES. To make thing simpler, we created two vectors female.logappearances and male.logappearances for you, which contains the log appearances counts for female and male characters separately. You will use these two vectors to do hypothesis testing in the rest of the lab.

```
female.logappearances <- log(marvel$APPEARANCES[marvel$SEX == "Female Characters"])
male.logappearances <- log(marvel$APPEARANCES[marvel$SEX == "Male Characters"])</pre>
```

T-test confidence intervals

Exercise 1.

(a) Get the confidence interval for the means of the female and male log appearance using the t.test() function.

```
# Insert your code here for calculating the CI, and save the CIs as
# `log.female.ci` and `log.male.ci`
log.female.ci <- t.test(female.logappearances)$conf.int
log.male.ci <- t.test(male.logappearances)$conf.int
log.female.ci
## [1] 1.556754 1.647675
## attr(,"conf.level")
## [1] 0.95
log.male.ci
## [1] -1.263904 -1.313981
## attr(,"conf.level")
## [1] 0.95</pre>
```

You may noticed that when you print the confidence interval, there is an additional line named attr(,"conf.level") with value 0.95. In R, all objects can have arbitrary additional attributes used to store metadata about the object. Attributes can be accessed using attributes() or attr(). For example, to get all the attributes of log.male.ci, run attributes(log.male.ci). And to get the conf.level attribute of log.male.ci, run attr(log.male.ci, "conf.level"). Attributes do not influence the fact that log.male.ci is a two-dimensional vector.

(b) Get the confidence interval for the difference in the log appearance count for female and male using the t.test() function.

```
# Insert your code here for calculating the CI, and save the CI as
# `log.diff.ci`
log.diff.ci <- - t.test(female.logappearances, male.logappearances)$conf.int
`log.diff.ci`
## [1] -0.2613763 -0.3651671
## attr(,"conf.level")
## [1] 0.95</pre>
```

- (c) Based on your calculation, which of the following statements do you support?
- i. There is no significant difference in appearance counts for male and female in Marvel comics.
- ii. The female character appearances count is significantly larger than the male in Marvel comics.
- iii. The male character appearances count is significantly larger than the female in Marvel comics.

```
# Uncomment the line of your answer for this question:
Ex1c.answer <- "ii"</pre>
```

Bootstrap confidence intervals

Exercise 2.

In this exercise, you will use the bootstrap to get the confidence interval for the difference in means between female and male log appearance count.

(a) Calculate the observed difference.

```
# Insert your code here for calculating the observed difference
obs.d <- t.test(female.logappearances, male.logappearances)$statistic
obs.d</pre>
```

t ## 11.83392

(b) Complete the following function to calculate the bootstrapped statistic. The function only needs to calculate one statistic; you will replicate it later.

You can test whether your code works or not by running the following chunk:

```
test <- bootOnce()
test # This should give you a numeric value, which is a bootstrapped statistic</pre>
```

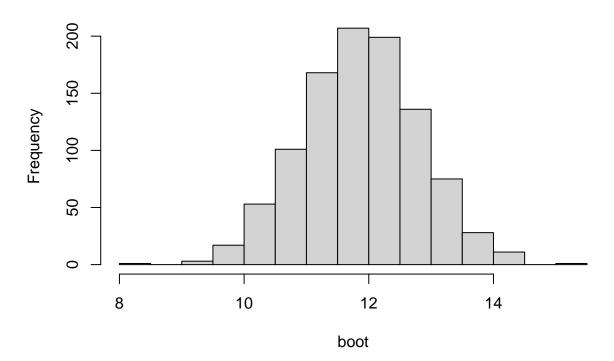
(c) Replicate the function bootOnce 1000 times to get 1000 bootstrapped statistics.

```
# Insert your code here and save the bootstrapped differences as
# `boot`
boot <- replicate(1000, bootOnce())</pre>
```

(d) Plot the histogram of your bootstrapped statistics.

```
# Insert your code for histogram here
hist(boot)
```

Histogram of boot



(e) Calculate the bootstrapped CI at the 0.95 confidence level. __Hint: __ use the quantile() function to find the bounds; use lower bound = (1 - 0.95)/2th quantile, and upper bound = 1 - (1 - 0.95)/2th quantile.

```
# Insert your code and save your bootstrapped CI as
# `boot.ci`
boot.ci <- - quantile(boot, probs = c(0.025, 0.975))</pre>
```

- (f) Based on your calculation, which of the following statements is supported?
- i. I will fail to reject the hypothesis that there is no significant difference in appearance counts for male and female.
- ii. I will reject the hypothesis that there is no significant difference in appearance counts for male and female.

```
# Uncomment the line of your answer for this question:
# Ex2f.answer <- "i."
Ex2f.answer <- "ii."</pre>
```

Linear regression

(Intercept)

dat\$x

```
Let us first look at a simulated example. You can ignore the particularities of the data generating process.
# Make some data
# X increases (noisily)
# Y is constructed so it is inversely related to xvar
set.seed(955)
xvar < -1:20 + rnorm(20, sd=3)
yvar \leftarrow -2 * xvar + 3 + rnorm(20, sd=4)
# Make a data frame with the variables
dat <- data.frame(x=xvar, y=yvar)</pre>
# Show first few rows
head(dat)
##
             Х
## 1 -4.252354 14.802251
## 2 1.702318 -3.063383
## 3 4.323054 -9.771777
## 4 1.780628 -3.130846
## 5 11.537348 -25.120776
## 6 6.672130 -13.533446
To do regression of y on x as a predictor, call the lm function:
# These two commands is equivalent
fit <- lm(y ~ x, data=dat)</pre>
fit <- lm(dat$y ~ dat$x)</pre>
To get detailed information about the fit, such as coefficient estimates, t-statistics and p-values:
summary(fit)
##
## Call:
## lm(formula = dat$y ~ dat$x)
## Residuals:
       Min
                 1Q Median
                                  ЗQ
                                         Max
## -5.0503 -2.6031 -0.2142 2.6217 7.6811
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 2.7791
                             1.4737
                                       1.886
                                               0.0756 .
                 -1.9889
                             0.1295 -15.355 8.71e-12 ***
## dat$x
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 3.642 on 18 degrees of freedom
## Multiple R-squared: 0.9291, Adjusted R-squared: 0.9251
## F-statistic: 235.8 on 1 and 18 DF, p-value: 8.709e-12
The coefficients estimation can be accessed:
fit$coefficients
```

2.779061 -1.988899

To get the estimated intercept term:

fit\$coefficients[1]

```
## (Intercept)
## 2.779061
```

To get the estimated slope term:

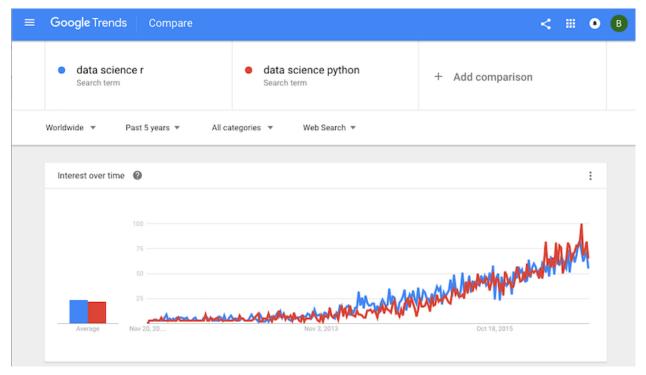
fit\$coefficients[2]

```
## dat$x
## -1.988899
```

Google Trends is a Google web tool providing equally-spaced time series data on the search volume. You may compare different topics to discover how peoples' interests change over time. You already code in at least one of the two most widely-used programming languages for data science (R and Python). Which language is more popular over time?

The following picture shows how to retrieve the Google Trends data. You may download and analyze your topics of interests as well. The dataset we obtained contains the following variables:

- week: beginning date of the week (recent 5 years)
- python: trend of the search term Data science Python
- r: trend of the search term Data science r



Read the data.

```
data_science <- read.csv("data_science.csv")
# convert string to date object
data_science$week <- as.Date(data_science$week, "%Y-%m-%d")
# create a numeric column representing the time</pre>
```

```
data_science$time <- as.numeric(data_science$week)
data_science$time <- data_science$time - data_science$time[1] + 1</pre>
```

Exercise 3

Run a regression of the ${\tt r}$ index on the predictor time. Get the estimate of the slope.

```
# Insert your code here and save the estimated slope as
# `r.slope`
r.slope <- lm(r ~ time, data = data_science)$coefficients[2]
r.slope
## time</pre>
```

time ## 0.03842291