# Homework 9

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2022-04-20

# Chapter 13

#### Problem 6

Sally and Joan plan to meet to study in their college campus center. They are both impatient people who will only wait 10 minutes for the other before leaving. Rather than pick a specific time to meet, they agree to head over to the campus center sometime between 7:00 and 8:00 pm. Let both arrival times be normally distributed with mean 30 minutes past and a standard deviation of 10 minutes. Assume that they are independent of each other. What is the probability that they actually meet? Estimate the answer using simulation techniques introduced in this chapter, with at least 10,000 simulations.

```
n <- 10000
sim_meet <- tibble::tibble(</pre>
    sally = rnorm(n, mean = 30, sd = 10),
    joan = rnorm(n, mean = 30, sd = 10),
    result = ifelse(
        abs(sally - joan) <= 10, "They meet", "They do not"
)
mosaic::tally(~ result, format = "percent", data = sim_meet)
## Registered S3 method overwritten by 'mosaic':
##
     method
                                       from
     fortify.SpatialPolygonsDataFrame ggplot2
##
## result
## They do not
                 They meet
##
         48.93
                     51.07
mosaic::binom.test(~result, n, success = "They meet", data = sim_meet)
##
##
##
## data: sim_meet$result [with success = They meet]
## number of successes = 5107, number of trials = 10000, p-value = 0.03317
## alternative hypothesis: true probability of success is not equal to 0.5
## 95 percent confidence interval:
## 0.5008505 0.5205433
```

```
## sample estimates:
## probability of success
## 0.5107
```

### Problem 9

Generate n = 5000 observations from a logistic regression model with parameters intercept  $\beta_0 = -1$ , slope  $\beta_1 = 0.5$ , and distribution of the predictor being normal with mean  $\mu = 1$  and standard deviation  $\sigma = 1$ . Calculate and interpret the resulting parameter estimates and confidence intervals.

```
set.seed(500)
# Generate 5000 predictor variables
x \leftarrow rnorm(n = 5000, mean = 1, sd = 1)
# Set beta values
beta_0 <- -1
beta 1 <- 0.5
# Generate set of probabilities
true_probs \leftarrow exp(beta_0 + beta_1 * x) / (1 + exp(beta_0 + beta_1 * x))
# Feed probabilities into a binomial function
y <- rbinom(n = 5000, size = 1, prob = true_probs)
# Generate data frame using predictor and probability values
sim_data \leftarrow dplyr::tibble(X = x, Y = y)
log_reg <- glm(Y ~ X, data = sim_data, family = "binomial")</pre>
log_summary <- summary(log_reg)</pre>
log_summary
##
## Call:
## glm(formula = Y ~ X, family = "binomial", data = sim_data)
## Deviance Residuals:
##
       Min
                 1Q
                      Median
                                    3Q
                                            Max
## -1.6941 -0.9747 -0.7874
                                1.2542
                                         2.0241
##
## Coefficients:
##
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.99669
                            0.04572 -21.80
                                              <2e-16 ***
                                              <2e-16 ***
## X
                0.49084
                            0.03173
                                      15.47
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 6644.5 on 4999 degrees of freedom
## Residual deviance: 6387.0 on 4998 degrees of freedom
## AIC: 6391
##
## Number of Fisher Scoring iterations: 4
```

```
# log_summary$coefficients holds the data frame for the coefficients.
# Column 1 is the estimated values, column 2 is the standard errors

beta_0_est <- log_summary$coefficients[1,1]
beta_0_error <- log_summary$coefficients[1,2]
beta_1_est <- log_summary$coefficients[2,1]
beta_1_error <- log_summary$coefficients[1,2]</pre>
```

```
# Create function to help calc intervals

calc_ci <- function(estimate, error) {
   low <- estimate - (2 * error)
   high <- estimate + (2 * error)

   c(low, high)
}</pre>
```

```
# Calculate confidence intervals and round
beta_0_ci <- calc_ci(beta_0_est, beta_0_error) %>% round(3)
beta_1_ci <- calc_ci(beta_1_est, beta_1_error) %>% round(3)
```

```
## The confidence interval for beta 0 is [-1.088, -0.905].
## The confidence interval for beta 1 is [0.399, 0.582].
```

The parameter estimates were largely spot on! The confidence intervals are relatively small and neatly capture the true parameter value.

# Chapter 19

#### Problem 3

**Note:** Copying and pasting these into this messes with the formatting of this page a bit. I'm not sure how to fix that!

#### Guesses:

- 1: str\_subset(x, pattern = "pop")
  - This will return the first 7 words in the vector, specifically because they all have "pop" in them.
- 2: str detect(x, pattern = "^pop")
  - This will return a vector of booleans, the first 6 being TRUE. The rest will be FALSE.
- 3: str detect(x, pattern = "populari[sz]e")
  - Index 3 and 4 will be TRUE, the rest of the vector will be FALSE.
- 4: str detect(x, pattern = "pop.\*e")
  - This looks for words containing some combination of [pop...e]. Indices 3, 4, 7 will be TRUE, the rest FALSE.
- 5: str detect(x, pattern = "p[a-z]\*e")
  - This looks for any number of any letter contained between a "p" and an "e". Indices 3, 4, 7, 8 will be TRUE, the rest FALSE (though im unsure of index 10 here).
- 6:  $str\_detect(x, pattern = "^[Pp][a-z]+.*n")$ 
  - This looks for words starting with either "P" or "p", followed by any letter. The plus indicates any number of letters, followed by any number of any type of character. Finally, we end with an n. This applies to indices 6 and 12. So those will be TRUE, the rest FALSE.
- 7: str subset(x, pattern = "^[^Pp]")
  - This just looks for words starting with anything BUT "P" or "p". So this returns indices 7, 8, 9 and 11.
- 8: str detect(x, pattern = "^[A-Za-p]")
  - This looks for words starting with either a capital letter or a lowercase letter. That applies to everything BUT index 10. So index 10 is TRUE, all the rest are false.
- 9: str\_detect(x, pattern = "[]")
  - This just looks for a space character which is found in indices 9, 10 and 11. Those three will be TRUE, the rest FALSE.
- 10: str\_subset(x, pattern = "[]")
  - This looks for a tab. One is found on index 10, so that is the only string returned by this.
- 11:  $str\_detect(x, pattern = "[]^{\hat{n}})$

- Same as last time, but this also wants a space before the tab. Nothing meets that criteria, so this
  will return a vector of only FALSE values.
- 12: str\_subset(x, pattern = "1")
  - This returns a vector containing any string starting with a space, which is index 11.

## Running commands:

```
x <- c("popular", "popularity", "popularize", "popularise", "Popular",
                                 "population", "repopulate", "reproduce", "happy family",
                                 "happier\tfamily", " happy family", "P6dn")
str_subset(x, pattern = "pop")
                                                                                             "popularity" "popularize" "popularise" "population"
## [1] "popular"
## [6] "repopulate"
str_detect(x, pattern = "^pop")
                [1] TRUE TRUE TRUE TRUE FALSE TRUE FALSE FALSE FALSE FALSE FALSE
str_detect(x, pattern = "populari[sz]e")
                 [1] FALSE FALSE TRUE TRUE FALSE FALS
str_detect(x, pattern = "pop.*e")
                 [1] FALSE FALSE TRUE TRUE FALSE FALS
str_detect(x, pattern = "p[a-z]*e")
                [1] FALSE FALSE TRUE TRUE FALSE FALSE TRUE TRUE FALSE TRUE FALSE FALSE
str_detect(x, pattern = "^[Pp][a-z]+.*n")
                [1] FALSE FALSE FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE
str_subset(x, pattern = "^[^Pp]")
## [1] "repopulate"
                                                                                                                    "reproduce"
                                                                                                                                                                                                        "happy family"
                                                                                                                                                                                                                                                                                           "happier\tfamily"
## [5] " happy family"
str_detect(x, pattern = "^[A-Za-p]")
               [1] TRUE TRUE TRUE TRUE TRUE TRUE FALSE FALSE TRUE TRUE FALSE TRUE
```

```
str_detect(x, pattern = "[]") #9

## [1] FALSE FALSE FALSE FALSE FALSE FALSE FALSE TRUE FALSE

str_subset(x, pattern = "[\t]") #10

## [1] "happier\tfamily"

str_detect(x, pattern = "[\t]") #11

## [1] FALSE FALSE FALSE FALSE FALSE FALSE FALSE TRUE TRUE TRUE TRUE FALSE

str_subset(x, pattern = "^[]") #12

## [1] " happy family"
```

## Problem 4

Use the babynames data table from the babynames package to find the 10 most popular:

```
a: Boys names ending in a vowel.b: Names ending with 'joe', 'jo', 'Joe', 'Jo' (e.g., Billyjoe).
```

```
# A
# Firstly I want to make sure I don't save babynames as a variable. It's HUGE, so we pipe
# it in directly. Secondly we filter using str_detect, putting all the vowels in brackets.
# Include sex requirement as well. From there you just group the names together and sum up
# the counts for all of the years.
# Sort in descending order and restrict to the top 10 and you're done!

babynames::babynames %>%
    filter(
        stringr::str_detect(string = name, pattern = "[AEIOUaeiou]$") & sex == "M"
        ) %>%
        group_by(name) %>%
        summarise(popularity = sum(n)) %>%
        arrange(desc(popularity)) %>%
        head(n = 10)
```

```
## 5 Lawrence 456773
           450780
448702
## 6 Joe
## 7 Willie
## 8 Jesse
                416530
## 9 Bruce
                 382257
## 10 Eugene
                378539
# B
babynames::babynames %>%
   filter(
       stringr::str_detect(string = name, pattern = "(joe|jo|Joe|Jo)$")
   ) %>%
   group_by(name) %>%
   summarise(popularity = sum(n)) %>%
   arrange(desc(popularity)) %>%
   head(n = 10)
```

```
## # A tibble: 10 x 2
## name popularity
            <int>
##
     <chr>
## 1 Joe
               462099
               180579
## 2 Jo
## 3 Maryjo
                  7017
## 4 Billiejo
                 1455
## 5 Marijo
                  1280
## 6 Bobbijo
                 1237
## 7 Bobbiejo
                 1009
## 8 Alejo
                   794
## 9 Bettyjo
                   764
## 10 Amyjo
                   486
```