

Homework 3 Answers

January 26, 2022

Instructions

- This homework is due Wednesday, February 2 at 3pm EST.
- Submit via GitHub. Remember to commit and push online so I can see it.
- Please format your homework solutions using R Markdown. You are welcome to simply add your answers below each question.
 - If the question requires a figure, make sure you have informative title, axis labels, and legend if needed.
 - Note: When I’ve given the framework of an answer’s code, I’ve included the option `eval=FALSE` in the R chunk. When you start filling in your answer, you’ll need to switch this to `eval=TRUE`.
- Turn in both the .rmd file and the knitted .pdf or .html file.
 - Knitting the .rmd file to a .pdf or .html file should help ensure your code runs without errors, but double check the output is what you expected.

Question 1

- Sampling variability
- Sampling distribution
- Standard deviation
- Standard error

For each of these terms, in your own words:

1. explain what it means in terms of \hat{ATE} .
2. explain how it relates to the other terms listed.

(Don’t be afraid of repeating yourself!)

Question 2

Read the Exercise 8 prompt from GG Ch. 3. Although a natural experiment, explain in your own words how this data was generated, in effect, using a block randomized design. In other words, why should we analyze this data using what we know about estimating \hat{ATE} and $SE(\hat{ATE})$ under block randomization?

Answer:

The term-length was chosen randomly for these state senators *within state*. Those are the two key ingredients—random assignment, but within each state separately. Therefore, it’s as if a researcher did a blocked design, with state being the blocks. We should analyze the data in light of the design i.e., how randomization took place!

Question 3

Complete Exercise 8 from GG Ch. 3, parts a, b, c, and e.

Note: do these questions “by hand,” meaning, calculate the quantities on your own in code without using pre-programmed statistical routines like `lm()`. (You can still use commands like `mean()`!)

In this code chunk, I’m just creating the dataset shown in the book for you.

```

term_length_tx <- c(rep(0, 16),
                    rep(1, 15))
term_length_ak <- c(rep(0, 16),
                    rep(1, 18), 0)
bills_tx <- c(18,29,41,53,60,67,75,79,79,
              88,93,101,103,106,107,131,
              29,37,42,45,45,54,54,58,61,
              64,69,73,75,92,104)
bills_ak <- c(11,15,23,24,25,26,28,31,33,
              34,35,35,36,38,52,59,9,
              10,14,15,15,17,18,19,19,
              20,21,23,23,24,28,30,32,34,17)
df <- data.frame(state = c(rep("TX", 31),
                           rep("AK", 35)),
                 term_length = c(term_length_tx, term_length_ak),
                 bills = c(bills_tx, bills_ak))

```

3a

Answer:

The \hat{ATE}_{AK} is:

```

ate_est_ak <- mean(df$bills[df$term_length == 1 & df$state == "AK"]) -
  mean(df$bills[df$term_length == 0 & df$state == "AK"])
ate_est_ak

```

```
## [1] -10.09477
```

The \hat{ATE}_{TX} is:

```

ate_est_tx <- mean(df$bills[df$term_length == 1 & df$state == "TX"]) -
  mean(df$bills[df$term_length == 0 & df$state == "TX"])
ate_est_tx

```

```
## [1] -16.74167
```

3b

Answer:

See equation 3.6 on page 61 of GG. The $\hat{SE}(\hat{ATE}_{AK})$ is:

```

y0 <- df$bills[df$term_length == 0 & df$state == "AK"]
y1 <- df$bills[df$term_length == 1 & df$state == "AK"]
se_est_ak <- sqrt(var(y0)/length(y0) + var(y1)/length(y1))
se_est_ak

```

```
## [1] 3.395979
```

The $\hat{SE}(\hat{ATE}_{TX})$ is:

```

y0 <- df$bills[df$term_length == 0 & df$state == "TX"]
y1 <- df$bills[df$term_length == 1 & df$state == "TX"]
se_est_tx <- sqrt(var(y0)/length(y0) + var(y1)/length(y1))
se_est_tx

```

```
## [1] 9.345871
```

3c

The overall \hat{ATE} is:

```
N_ak <- sum(df$state == "AK")
N_tx <- sum(df$state == "TX")
N <- nrow(df)

overall_ate_est <- (N_ak/N)*ate_est_ak + (N_tx/N)*ate_est_tx
overall_ate_est

## [1] -13.2168
```

3e

Answer:

See equation 3.12 on page 74 of GG. The $\hat{SE}(\hat{ATE})$ is:

```
sqrt(se_est_ak^2 * (N_ak/N)^2 + se_est_tx^2 * (N_tx/N)^2)

## [1] 4.74478
```

Question 4

Now, answer Exercise 8 part d.

In addition to explaining what the question asks you to explain, calculate \hat{ATE} when pooling and compare the estimated ATE to when you didn't pool in 3c.

Also, calculate $\hat{SE}(\hat{ATE})$ when pooling and compare the estimated ATE to when you didn't pool in 3e.

What are the implications if the researcher incorrectly analyzes this data?

Answer:

The two states differ in terms of the probability that a given legislator will be assigned to the treatment. Therefore, we cannot pool the data without introducing a correlation between treatment assignment and the potential outcomes associated with the two states. In this study, the experiments take place within each state, and the analyst should pool the state-level results in order to obtain an overall result.

```
overall_ate_est_pooled <- (mean(df$bills[df$term_length == 1]) - mean(df$bills[df$term_length == 0]))
overall_ate_est_pooled

## [1] -14.51515

y0 <- df$bills[df$term_length == 0]
y1 <- df$bills[df$term_length == 1]

overall_se_est_pooled <- sqrt(var(y0)/length(y0) + var(y1)/length(y1))
overall_se_est_pooled

## [1] 7.132193
```

Checking your answers

Confirm your correct results from Question 3 and your incorrect results from Question 4 using software commands available in R. I've provided the commands for you. If what you calculated in Question 3 and Question 4 matches up, then you know what's happening under the hood in these functions!

```

# with blocking
estimatr::difference_in_means(bills ~ term_length,
                              blocks = state,
                              data = df)

## Design: Blocked
##           Estimate Std. Error  t value    Pr(>|t|)  CI Lower  CI Upper DF
## term_length -13.2168    4.74478 -2.785545 0.007079688 -22.70148 -3.732117 62

# without blocking
estimatr::difference_in_means(bills ~ term_length,
                              data = df)

## Design: Standard
##           Estimate Std. Error  t value    Pr(>|t|)  CI Lower  CI Upper
## term_length -14.51515    7.132193 -2.03516 0.0463009 -28.78439 -0.245916
##           DF
## term_length 59.44775

```