# Problem Set 2

### Applied Stats/Quant Methods 1

Due: October 14, 2024

### Instructions

- Please show your work! You may lose points by simply writing in the answer. If the problem requires you to execute commands in R, please include the code you used to get your answers. Please also include the .R file that contains your code. If you are not sure if work needs to be shown for a particular problem, please ask.
- Your homework should be submitted electronically on GitHub.
- This problem set is due before 23:59 on Monday October 14, 2024. No late assignments will be accepted.

## **Question 1: Political Science**

The following table was created using the data from a study run in a major Latin American city.<sup>1</sup> As part of the experimental treatment in the study, one employee of the research team was chosen to make illegal left turns across traffic to draw the attention of the police officers on shift. Two employee drivers were upper class, two were lower class drivers, and the identity of the driver was randomly assigned per encounter. The researchers were interested in whether officers were more or less likely to solicit a bribe from drivers depending on their class (officers use phrases like, "We can solve this the easy way" to draw a bribe). The table below shows the resulting data.

	Not Stopped	Bribe requested	Stopped/given warning
Upper class	14	6	7
Lower class	7	7	1

<sup>&</sup>lt;sup>1</sup>Fried, Lagunes, and Venkataramani (2010). "Corruption and Inequality at the Crossroad: A Multimethod Study of Bribery and Discrimination in Latin America. *Latin American Research Review*. 45 (1): 76-97.

(a) Calculate the  $\chi^2$  test statistic by hand/manually (even better if you can do "by hand" in R).

Answer: The chi-squared test statistic is 3.791168, df = 2

```
1 #####1.1#####
2 # create data
3 class_bribery \leftarrow matrix (c(14,6,7,7,7,1), nrow = 2, ncol = 3, byrow = TRUE
4 # rename
rownames(class_bribery) <- c("Upper class", "Lower class")
colnames(class_bribery) <- c("Not Stopped", "Bribe requested", " Stopped/</pre>
      given warning")
8 # sum
9 sum_class_bribery <- cbind(class_bribery, row_sum = rowSums(class_bribery</pre>
10 sum_class_bribery <- rbind(sum_class_bribery, col_sum = colSums(sum_class_
      bribery))
12 # calculate frequency
13 expected_values <- class_bribery*0</pre>
14 for (i in 1:2) {
    for (j in 1:3) {
15
       expected_values[i,j] <- (sum_class_bribery[i, "row_sum"]*sum_class_
16
      bribery [-(1:2),j] /sum (class_bribery)
17
18 }
19
20 # chi-squared statistic
21 chi_statistic <- sum((class_bribery-expected_values)^2/expected_values)
print(chi_statistic)
23
24 # df
_{25} df \leftarrow (nrow(class_bribery)-1) * (ncol(class_bribery)-1)
26 print (df)
27
28 # test
chisq.test(class_bribery)
  > print(chi_statistic)
  [1] 3.791168
  > print(df)
  [1] 2
```

(b) Now calculate the p-value from the test statistic you just created (in R).<sup>2</sup> What do you conclude if  $\alpha = 0.1$ ?

Answer: Since the p-value (0.15) > 0.10, we don't have sufficient evidence to reject the null hypothesis at the alpha = 0.1 significance level. This suggests that there is no significant difference in the likelihood of police officers soliciting a bribe from drivers based on their class.

```
> print(pvalue)
[1] 0.1502306
```

(c) Calculate the standardized residuals for each cell and put them in the table below.

<sup>&</sup>lt;sup>2</sup>Remember frequency should be > 5 for all cells, but let's calculate the p-value here anyway.

	Not Stopped	Bribe requested	Stopped/given warning
Upper class	0.322	-1.642	1.523
Lower class	-0.322	1.642	-1.523

(d) How might the standardized residuals help you interpret the results?

#### Answer:

- 1.Standardized residuals with absolute values greater than 2 or less than -2 are typically considered outliers. According to the table, no observations show significant deviations from the model predictions.
- 2. The standardized residual is very small, which to some extent suggests that there is no dependency relationship present.
- $3.To\ draw\ conclusions$ , it is also necessary to consider the resualt of chi-squared test and the p-value.

## Question 2: Economics

Chattopadhyay and Duflo were interested in whether women promote different policies than men.<sup>3</sup> Answering this question with observational data is pretty difficult due to potential confounding problems (e.g. the districts that choose female politicians are likely to systematically differ in other aspects too). Hence, they exploit a randomized policy experiment in India, where since the mid-1990s,  $\frac{1}{3}$  of village council heads have been randomly reserved for women. A subset of the data from West Bengal can be found at the following link: https://raw.githubusercontent.com/kosukeimai/qss/master/PREDICTION/women.csv

Each observation in the data set represents a village and there are two villages associated with one GP (i.e. a level of government is called "GP"). Figure 1 below shows the names and descriptions of the variables in the dataset. The authors hypothesize that female politicians are more likely to support policies female voters want. Researchers found that more women complain about the quality of drinking water than men. You need to estimate the effect of the reservation policy on the number of new or repaired drinking water facilities in the villages.

Figure 1: Names and description of variables from Chattopadhyay and Duflo (2004).

$_{ m Name}$	Description		
GP	An identifier for the Gram Panchayat (GP)		
village	identifier for each village		
reserved	binary variable indicating whether the GP was reserved		
	for women leaders or not		
female	binary variable indicating whether the GP had a female		
	leader or not		
irrigation	variable measuring the number of new or repaired ir-		
	rigation facilities in the village since the reserve policy		
	started		
water	variable measuring the number of new or repaired		
	drinking-water facilities in the village since the reserve		
	policy started		

<sup>&</sup>lt;sup>3</sup>Chattopadhyay and Duflo. (2004). "Women as Policy Makers: Evidence from a Randomized Policy Experiment in India. *Econometrica*. 72 (5), 1409-1443.

(a) State a null and alternative (two-tailed) hypothesis.

#### Answer:

Ho = Having reserved village council heads for female does not affect the number drinking water facilities in the villages.

Ha = Having reserved village council heads for female affects the number drinking water facilities in the villages.

(b) Run a bivariate regression to test this hypothesis in R (include your code!).

```
1 ######2. 2#######
2 data_problem2 <- read.csv("https://raw.githubusercontent.com/kosukeimai/
     qss/master/PREDICTION/women.csv", header=T)
3 regression_model <- lm(water ~ reserved, data = data_problem2)</pre>
4 summary (regression _model)
 >summary(regression_model)
 Call:
 lm(formula = water ~ reserved, data = data_problem2)
 Residuals:
 Min
          10 Median
                           3Q
                                  Max
 -23.991 -14.738 -7.865
                          2.262 316.009
 Coefficients:
 Estimate Std. Error t value Pr(>|t|)
               14.738
                            2.286
 (Intercept)
                                    6.446 4.22e-10 ***
 reserved
                9.252
                            3.948
                                    2.344
                                            0.0197 *
 Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' '1
 Residual standard error: 33.45 on 320 degrees of freedom
 Multiple R-squared: 0.01688, Adjusted R-squared: 0.0138
 F-statistic: 5.493 on 1 and 320 DF, p-value: 0.0197
```

(c) Interpret the coefficient estimate for reservation policy.

#### Answer:

Since the p-value (approximately 0.02) < 0.05, we have sufficient evidence to reject the null hypothesis at the alpha = 0.05 significance level. This suggests that Having reserved seats for female politicians increase the number drinking water facilities in the villages.