Problem Set 3

Applied Stats/Quant Methods 1

Due: November 19, 2022

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 Sunday November 19, 2023. No late assignments will be accepted.

In this problem set, you will run several regressions and create an add variable plot (see the lecture slides) in R using the incumbents_subset.csv dataset. Include all of your code.

Question 1

We are interested in knowing how the difference in campaign spending between incumbent and challenger affects the incumbent's vote share.

1. Run a regression where the outcome variable is **voteshare** and the explanatory variable is **difflog**.

```
# This analyse will be about a dataset called incumbents subset.csv
inc.sub <- read.csv("https://raw.githubusercontent.com/ASDS-TCD/StatsI_Fall2023/main/datasets/incumbents_subset.csv")

# Check out the head and tail.
head(inc.sub)

# We will select only the columns "voteshare" and "difflog" to create a dataframe with the relevant variables
selected_data <- inc.sub[, c("voteshare", "difflog")]

# Simple linear regression
model <- lm(voteshare ~ difflog, data = selected_data)

# Summary of the regression model
summary(model)
```

```
Call:
lm(formula = voteshare ~ difflog, data = selected_data)
Residuals:
     Min
               10
                  Median
                                30
                                        Max
-0.26832 -0.05345 -0.00377 0.04780 0.32749
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                257.19
                                         <2e-16 ***
(Intercept) 0.579031
                      0.002251
                                         <2e-16 ***
           0.041666
                      0.000968
                                43.04
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '. 0.1 ' 1
Residual standard error: 0.07867 on 3191 degrees of freedom
Multiple R-squared: 0.3673,
                              Adjusted R-squared: 0.3671
F-statistic: 1853 on 1 and 3191 DF, p-value: < 2.2e-16
```

2. Make a scatterplot of the two variables and add the regression line.

```
plot(inc.sub$difflog, inc.sub$voteshare)
abline(model, col = "red")
```

```
call:
lm(formula = voteshare ~ difflog + presvote, data = selected_data)
Residuals:
    Min
                  Median
              1Q
                                        Max
                                3Q
-0.25928 -0.04737 -0.00121 0.04618 0.33126
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                          <2e-16 ***
(Intercept) 0.4486442 0.0063297
                                  70.88
difflog
           0.0355431 0.0009455
                                  37.59
                                          <2e-16 ***
                                          <2e-16 ***
presvote
           0.2568770 0.0117637
                                  21.84
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.07339 on 3190 degrees of freedom
                              Adjusted R-squared: 0.4493
Multiple R-squared: 0.4496,
F-statistic: 1303 on 2 and 3190 DF, p-value: < 2.2e-16
```

3. Save the residuals of the model in a separate object.

```
residuals <- resid(model)

# Summary of the regression model
summary(model)

# Residuals
head(residuals)
```

We are interested in knowing how the difference between incumbent and challenger's spending and the vote share of the presidential candidate of the incumbent's party are related.

1. Run a regression where the outcome variable is **presvote** and the explanatory variable is difflog.

```
# We will select only the variables "presvote" and "difflog"
selected_data <- inc.sub [, c("presvote", "difflog")]
# Linear regression model
model_presvote <- lm(presvote ~ difflog, data = inc.sub)
# Summary
summary(model_presvote)</pre>
```

```
lm(formula = presvote ~ difflog, data = inc.sub)
Residuals:
                   Median
              1Q
                                3Q
-0.32196 -0.07407 -0.00102 0.07151 0.42743
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.507583 0.003161 160.60
difflog
           0.023837
                      0.001359
                                 17.54
                                         <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.1104 on 3191 degrees of freedom
Multiple R-squared: 0.08795, Adjusted R-squared: 0.08767
F-statistic: 307.7 on 1 and 3191 DF, p-value: < 2.2e-16
```

2. Make a scatterplot of the two variables and add the regression line.

```
plot(inc.sub$difflog, inc.sub$presvote)
# Regression
bline(model_presvote, col = "red")

plot(inc.sub$difflog, inc.sub$presvote)
red")
```

```
call:
lm(formula = voteshare ~ difflog + presvote, data = selected_data)
Residuals:
    Min
                   Median
              1Q
                                        Max
                                3Q
-0.25928 -0.04737 -0.00121 0.04618 0.33126
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                          <2e-16 ***
(Intercept) 0.4486442 0.0063297
                                  70.88
difflog
           0.0355431 0.0009455
                                  37.59
                                          <2e-16 ***
presvote
           0.2568770 0.0117637
                                  21.84
                                          <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.07339 on 3190 degrees of freedom
                              Adjusted R-squared: 0.4493
Multiple R-squared: 0.4496,
F-statistic: 1303 on 2 and 3190 DF, p-value: < 2.2e-16
```

3. Save the residuals of the model in a separate object.

```
# Residuals
residuals_presvote <- resid(model_presvote)

# Summary of the regression model
summary(model_presvote)

# Residuals
head(residuals_presvote)</pre>
```

We are interested in knowing how the vote share of the presidential candidate of the incumbent's party is associated with the incumbent's electoral success.

1. Run a regression where the outcome variable is **voteshare** and the explanatory variable is **presvote**.

```
# We will select the columns 'voteshare' and 'presvote'
selected_data <- inc.sub[, c("voteshare", "presvote")]

# Linear regression model
model_voteshare <- lm(voteshare ~ presvote, data = selected_data)

# Summary of the regression model
summary(model_voteshare)</pre>
```

```
lm(formula = voteshare ~ presvote, data = selected_data)
Residuals:
    Min
                   Median
               1Q
                                        Max
-0.27330 -0.05888 0.00394 0.06148 0.41365
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                         <2e-16 ***
                                  58.08
(Intercept) 0.441330
                      0.007599
                                          <2e-16 ***
presvote
           0.388018
                      0.013493
                                  28.76
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.08815 on 3191 degrees of freedom
Multiple R-squared: 0.2058,
                               Adjusted R-squared: 0.2056
F-statistic:
              827 on 1 and 3191 DF, p-value: < 2.2e-16
```

2. Make a scatterplot of the two variables and add the regression line.

```
# Scatterplot
plot(selected_data$presvote, selected_data$voteshare)

# Regression
bline(model_voteshare, col = "red")
```

```
call:
lm(formula = voteshare ~ difflog + presvote, data = selected_data)
Residuals:
              1Q Median
    Min
                                3Q
                                       Max
-0.25928 -0.04737 -0.00121 0.04618 0.33126
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                         <2e-16 ***
(Intercept) 0.4486442 0.0063297
                                  70.88
                                         <2e-16 ***
difflog
           0.0355431 0.0009455
                                  37.59
presvote
           0.2568770 0.0117637
                                  21.84
                                         <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.07339 on 3190 degrees of freedom
Multiple R-squared: 0.4496, Adjusted R-squared: 0.4493
F-statistic: 1303 on 2 and 3190 DF, p-value: < 2.2e-16
```

The residuals from part (a) tell us how much of the variation in **voteshare** is *not* explained by the difference in spending between incumbent and challenger. The residuals in part (b) tell us how much of the variation in **presvote** is *not* explained by the difference in spending between incumbent and challenger in the district.

1. Run a regression where the outcome variable is the residuals from Question 1 and the explanatory variable is the residuals from Question 2.

```
# 4 The residuals from Question 1 are stored in residuals and residuals
    from Question 2 are stored in residuals_presvote

residuals_model <- lm(residuals residuals_presvote)

# Summary of the regression model
summary(residuals_model)</pre>
```

```
call:
lm(formula = residuals ~ residuals_presvote)
Residuals:
     Min
               1Q
                    Median
                                         Max
-0.25928 -0.04737 -0.00121
                            0.04618
Coefficients:
                     Estimate Std. Error t value Pr(>|t|)
                   -5.934e-18 1.299e-03
                                            0.00
(Intercept)
                                                   <2e-16 ***
residuals_presvote 2.569e-01 1.176e-02
                                           21.84
                0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
Residual standard error: 0.07338 on 3191 degrees of freedom
                      0.13,
Multiple R-squared:
                                Adjusted R-squared:
F-statistic:
               477 on 1 and 3191 DF,
                                      p-value: < 2.2e-16
```

2. Make a scatterplot of the two residuals and add the regression line.

```
# Scatterplot with regression line
plot(residuals_presvote, residuals)
abline(residuals_model, col = "red")
```

```
call:
lm(formula = voteshare ~ difflog + presvote, data = selected_data)
Residuals:
              1Q Median
    Min
                                3Q
                                       Max
-0.25928 -0.04737 -0.00121 0.04618 0.33126
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                         <2e-16 ***
(Intercept) 0.4486442 0.0063297
                                  70.88
                                         <2e-16 ***
difflog
           0.0355431 0.0009455
                                  37.59
presvote
           0.2568770 0.0117637
                                  21.84
                                         <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.07339 on 3190 degrees of freedom
Multiple R-squared: 0.4496, Adjusted R-squared: 0.4493
F-statistic: 1303 on 2 and 3190 DF, p-value: < 2.2e-16
```

What if the incumbent's vote share is affected by both the president's popularity and the difference in spending between incumbent and challenger?

1. Run a regression where the outcome variable is the incumbent's voteshare and the explanatory variables are difflog and presvote.

```
# We will select the columns 'voteshare', 'difflog', and 'presvote'
selected_data <- inc.sub[, c("voteshare", "difflog", "presvote")]

# Linear regression model
model_combined <- lm(voteshare ~ difflog + presvote, data = selected_data
)

# Summary of the regression model
summary(model_combined)</pre>
```

```
lm(formula = voteshare ~ difflog + presvote, data = selected_data)
Residuals:
    Min
              1Q
                   Median
                                 3Q
                                        Max
-0.25928 -0.04737 -0.00121 0.04618 0.33126
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                          <2e-16 ***
(Intercept) 0.4486442 0.0063297
                                  70.88
           0.0355431 0.0009455
                                          <2e-16 ***
difflog
                                  37.59
presvote
           0.2568770 0.0117637
                                  21.84
                                          <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.07339 on 3190 degrees of freedom
Multiple R-squared: 0.4496,
                              Adjusted R-squared: 0.4493
F-statistic: 1303 on 2 and 3190 DF, p-value: < 2.2e-16
```

2. Write the prediction equation.

What is it in this output that is identical to the output in Question 4? Why do you think this is the case?