# Problem Set 3

# Applied Stats/Quant Methods 1

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## 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**.

```
#Question1:
model1 <- lm(voteshare ~ difflog, data = inc.sub)

# Summary of the regression model
summary(model1)</pre>
```

#### Result:

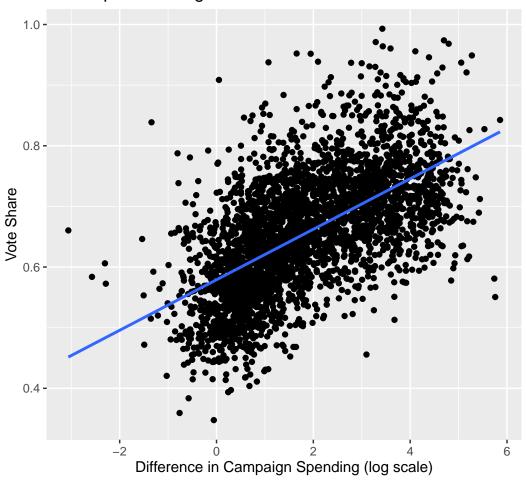
```
Call:
lm(formula = voteshare ~ difflog, data = data)
Residuals:
     Min
               1Q
                   Median
                                3Q
                                        Max
-0.26832 -0.05345 -0.00377 0.04780 0.32749
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.579031
                      0.002251 257.19
                                         <2e-16 ***
           0.041666
                      0.000968
                                 43.04
                                         <2e-16 ***
difflog
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.

```
# Create a scatterplot of difflog and voteshare
ggplot(inc.sub, aes(x = difflog, y = voteshare)) +
geom_point() +
geom_smooth(method = "lm", se = FALSE) +
labs(title = "Scatterplot of difflog vs. voteshare",
x = "Difference in Campaign Spending (log scale)",
y = "Vote Share")
```

### Result:

## Scatterplot of difflog vs. voteshare



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

```
# Save residuals of the model
residuals <- resid(model1)
write.csv(data.frame(residuals), "residuals.csv")
```

4. Write the prediction equation.

```
# Getting the coefficients of the regression model
coefficients <- coef(model1)

# Extract the intercept and slope
intercept <- coefficients[1]
slope <- coefficients[2]

# Output prediction equation
```

```
9 cat("prediction equation: voteshare =", round(intercept, 2), "+", round(
     slope, 2), "* difflog\n")
11 #If the lm function is not used, it can be explained in detail with the
      following procedure.
12 # Extraction of relevant variables
13 x <- inc.sub$difflog
14 y <- inc.sub$voteshare
16 # Calculate the mean value
17 \text{ x} - \text{mean} \leftarrow \text{mean}(x)
y_mean < mean(y)
20 # Calculation of regression coefficients
_{21} beta_1 <- sum((x - x_mean) * (y - y_mean)) / sum((x - x_mean)^2)
y_m = 0 < y_m = 0 + y_m = 0
24 # Print Factor
25 cat("ratio (beta_1):", beta_1, "\n")
26 cat("intercept (beta_0):", beta_0, "\n")
27 cat ("prediction equation: voteshare =", round(beta_0, 2), "+", round(beta
  _{-1}, 2), "* difflog \n")
```

#### Result:

prediction equation: voteshare = 0.58 + 0.04 \* difflog ratio (beta-1): 0.04166632 intercept (beta-0): 0.5790307 prediction equation: voteshare = 0.58 + 0.04 \* difflog

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.

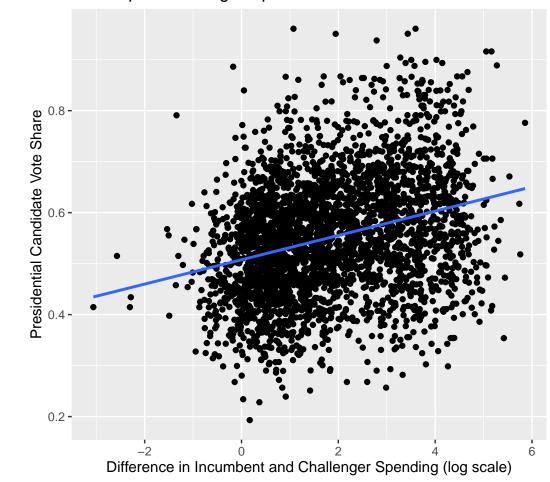
```
1 # Run a regression where the outcome variable is presvote and the
     explanatory variable is difflog
2 model2 <- lm(presvote ~ difflog, data = inc.sub)
4 # Summary of the regression model
5 summary (model2)
 Result:
 call:
  lm(formula = presvote ~ difflog, data = inc.sub)
  Residuals:
                     Median
      Min
                1Q
                                   3Q
                                          Max
  -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
                        0.001359
                                            <2e-16 ***
                                   17.54
             0.023837
  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.

```
# Create a scatterplot with a regression line
ggplot(inc.sub, aes(x = difflog, y = presvote)) +
geom_point() +
geom_smooth(method = "lm", se = FALSE) +
labs(title = "Scatterplot of difflog vs. presvote",
x = "Difference in Incumbent and Challenger Spending (log scale)",
y = "Presidential Candidate Vote Share")
```

#### Result:

### Scatterplot of difflog vs. presvote



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

```
# Save residuals of the model
residuals <- resid (model2)
```

4. Write the prediction equation.

```
# Extract the intercept and slope
coefficients <- coef(model2)
intercept <- coefficients[1]
slope <- coefficients[2]

# Output prediction equation
cat("prediction equation: presvote =", round(intercept, 2), "+", round(slope, 2), "* difflog\n")</pre>
```

#### Result:

prediction equation: presvote = 0.51 + 0.02 \* difflog

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**.

```
1 # Run a regression where the outcome variable is voteshare and the
     explanatory variable is presvote
2 model3 <- lm(voteshare ~ presvote, data = inc.sub)
4 # Summary of the regression model
5 summary (model3)
 Result:
  call:
  lm(formula = voteshare ~ presvote, data = inc.sub)
  Residuals:
                      Median
       Min
                 1Q
                                   3Q
                                           Max
  -0.27330 -0.05888 0.00394 0.06148
                                      0.41365
  Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                                            <2e-16 ***
  (Intercept) 0.441330 0.007599
                                    58.08
              0.388018
                         0.013493
                                    28.76
                                            <2e-16 ***
  presvote
  Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
  Residual standard error: 0.08815 on 3191 degrees of freedom
                               Adjusted R-squared: 0.2056
  Multiple R-squared: 0.2058,
                 827 on 1 and 3191 DF, p-value: < 2.2e-16
```

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

```
# Make a scatterplot of the two variables and add the regression line
ggplot(inc.sub, aes(x = presvote, y = voteshare)) +

geom_point() +

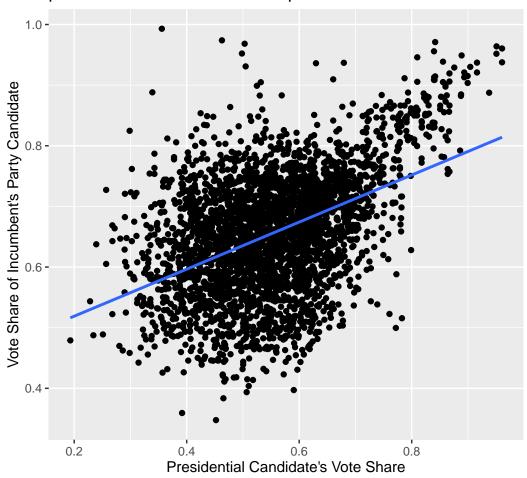
geom_smooth(method = "lm", se = FALSE) +

labs(title = "presvote vs. voteshare Scatterplot",

x = "Presidential Candidate's Vote Share",
y = "Vote Share of Incumbent's Party Candidate")
```

### Result:

## presvote vs. voteshare Scatterplot



### 3. Write the prediction equation.

```
# Extract coefficients
coefficients <- coef(model3)
intercept <- coefficients[1]
slope <- coefficients[2]

# Write prediction equation
cat("prediction equation: voteshare =", round(intercept, 2), "+", round(slope, 2), "* presvote\n")</pre>
```

### Result:

prediction equation: voteshare = 0.44 + 0.39 \* presvote

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.

```
1 # Extract residuals from the first regression
2 residuals1 <- resid (model1)</pre>
4 # Extract residuals from the second regression
5 residuals 2 <- resid (model 2)
7 # Run a regression where the outcome variable is the residuals from
      Question 1 and the explanatory variable is the residuals from Question
8 \mod 14 \leftarrow \ln (\operatorname{residuals2} \sim \operatorname{residuals1})
10 # Summary of the regression model
11 summary (model4)
  Result:
  call:
  lm(formula = residuals2 ~ residuals1)
  Residuals:
       Min
                       Median
                  1Q
                                     3Q
                                              Max
  -0.37076 -0.07095 0.00381 0.07404 0.30569
  Coefficients:
                Estimate Std. Error t value Pr(>|t|)
  (Intercept) 1.361e-18 1.823e-03
                                        0.00
                                                     1
  residuals1 5.062e-01 2.318e-02
                                        21.84
                                                <2e-16 ***
  signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
  Residual standard error: 0.103 on 3191 degrees of freedom
  Multiple R-squared:
                         0.13,
                                    Adjusted R-squared: 0.1298
                  477 on 1 and 3191 DF, p-value: < 2.2e-16
  F-statistic:
```

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

```
# Scatterplot with regression line for residuals
ggplot(data.frame(residuals1, residuals2), aes(x = residuals1, y =
    residuals2)) +
geom_point() +
geom_smooth(method = "lm", se = FALSE) +
```

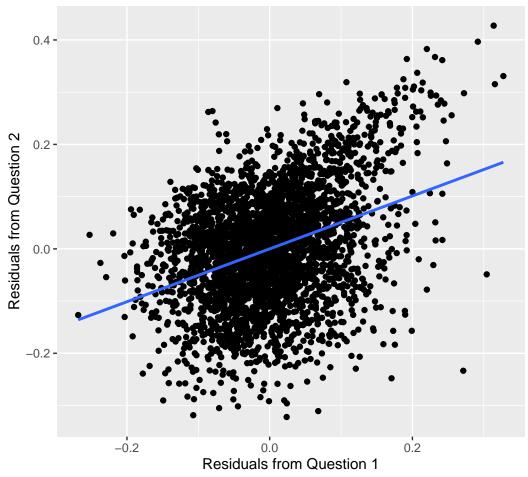
```
labs(title = "Residuals Scatterplot",

x = "Residuals from Question 1",

y = "Residuals from Question 2")
```

### Result:

# Residuals Scatterplot



### 3. Write the prediction equation.

```
# Extract coefficients for the regression with residuals
coefficients_residuals <- coef(model4)
intercept_residuals <- coefficients_residuals[1]
slope_residuals <- coefficients_residuals[2]

# Write prediction equation for residuals
cat("Prediction equation for residuals: residuals2 =", round(intercept_residuals, 2), "+", round(slope_residuals, 2), "* residuals1\n")
```

#### Result:

Prediction equation for residuals: residuals2 = 0 + 0.51 \* residuals1

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.

```
1 # Run a regression where the outcome variable is the incumbent's
     voteshare and the explanatory variables are difflog and presvote.
2 model_multiple <- lm(voteshare ~ difflog + presvote, data = inc.sub)
4 # Summary of the regression model
5 summary (model_multiple)
 Result:
 call:
 lm(formula = voteshare ~ difflog + presvote, data = inc.sub)
 Residuals:
                     Median
      Min
                1Q
                                  30
                                          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 '**' 0.05 '.' 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.

```
# Extract coefficients for the multiple regression
coefficients_multiple <- coef(model_multiple)
intercept_multiple <- coefficients_multiple[1]
slope_difflog <- coefficients_multiple[2]
slope_presvote <- coefficients_multiple[3]

# Write prediction equation for multiple regression
cat("Prediction equation for multiple regression: voteshare =", round(intercept_multiple, 2), "+", round(slope_difflog, 2), "* difflog +", round(slope_presvote, 2), "* presvote\n")</pre>
```

#### Result:

Prediction equation for multiple regression: voteshare = 0.45 + 0.04 \* difflog + 0.26 \* presvote

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

### Result:

Both questions 4 and 5 see the incumbent's share of the vote affected by and the difference in spending between the incumbent and the challenger. And there is a linear relationship, with more spending leading to a higher vote share. But as can also be seen in question 5, the incumbent's share of the vote is affected by both presidential popularity and the difference in spending between the incumbent and the challenger.