Answer Key: Problem Set 3

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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 in .pdf form.
- 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.

```
# read in data
inc.sub <- read.csv("https://raw.githubusercontent.com/ASDS-TCD/StatsI_
Fall2023/main/datasets/incumbents_subset.csv")

# (a) run regression where the outcome variable is \texttt{voteshare}

# and the explanatory variable is \texttt{difflog}
model_a <- lm(voteshare ~ difflog, data= inc.sub)

# texreg(list(model_a), digits=3)</pre>
```

Table 1: Outcome variable is voteshare and the explanatory variable is difflog.

	Model 1	
(Intercept)	0.579^{***}	
	(0.002)	
$\operatorname{difflog}$	0.042^{***}	
	(0.001)	
\mathbb{R}^2	0.367	
$Adj. R^2$	0.367	
Num. obs.	3193	
RMSE	0.079	
***n < 0.001 **n < 0.01 *n < 0.05		

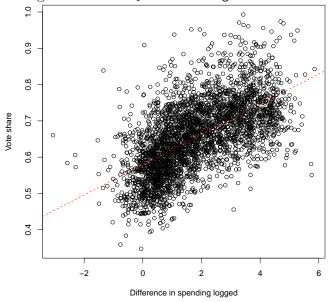
p < 0.001, **p < 0.01, *p < 0.05

There is a positive and statistically reliable relationship between the amount of spending between the incumbent and challenger and the incumbent's vote share, such that a one unit increase in the logged difference in spending is associated with an average increase of 0.04 in the incumbent's vote share (4%)

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

```
plot(inc.sub$difflog, inc.sub$voteshare,
      ylab="Vote share", xlab="Difference in spending logged")
abline (model_a, col="red", lty=2)
```

Figure 1: Scatter plot of difflog and voteshare.



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

```
model_a.resid <- resid(model_a)
```

4. Write the prediction equation.

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 \times \text{difflog}$$

 $voteshare = 0.579 + 0.042 \times \text{difflog}$

Question 2

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.

```
# (a) run regression where the outcome variable is \texttt{presvote}
# and the explanatory variable is \texttt{difflog}
model_b <- lm(presvote ~ difflog, data= inc.sub)</pre>
```

Table 2: Outcome variable is presvote and the explanatory variable is difflog.

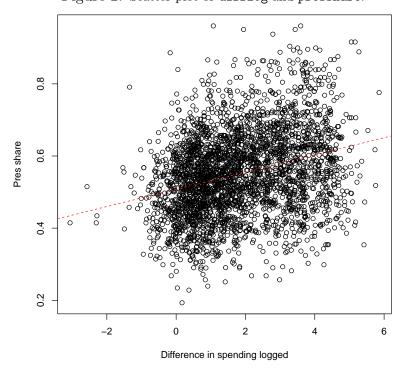
	Model 1
(Intercept)	0.508***
	(0.003)
$\operatorname{difflog}$	0.024^{***}
	(0.001)
\mathbb{R}^2	0.088
$Adj. R^2$	0.088
Num. obs.	3193
RMSE	0.110

^{***}p < 0.001, **p < 0.01, *p < 0.05

There is a positive and statistically reliable relationship between the amount of spending between the incumbent and challenger and the president vote share, such that a one unit increase in the logged difference in spending is associated with an average increase of 0.024 in the incumbent's vote share (2.4%)

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

Figure 2: Scatter plot of difflog and presshare.



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

```
1 model_b.resid <- resid (model_b)</pre>
```

4. Write the prediction equation.

$$presvote = 0.508 + 0.024 \times difflog$$

Question 3

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 # (a) run regression where the outcome variable is \texttt{voteshare}
2 # and the explanatory variable is \texttt{presvote}
3 model_c <- lm(voteshare ~ presvote, data= inc.sub)</pre>
```

Table 3: Outcome variable is voteshare and the explanatory variable is presvote.

	Model 1
(Intercept)	0.441***
	(0.008)
presvote	0.388***
	(0.013)
\mathbb{R}^2	0.206
$Adj. R^2$	0.206
Num. obs.	3193
RMSE	0.088
	0.01 * 0.05

^{***}p < 0.001, **p < 0.01, *p < 0.05

There is a positive and statistically reliable relationship between president vote share and voteshare, such that a one unit increase in the president's vote share is associated with an average increase of 0.388 in the incumbent's vote share (38.8%)

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

```
plot(inc.sub$presvote, inc.sub$voteshare,
ylab="Vote share", xlab="Pres share")
abline(model_c, col="red", lty=2)
```

3. Write the prediction equation.

 $voteshare = 0.441 + 0.388 \times presvote$

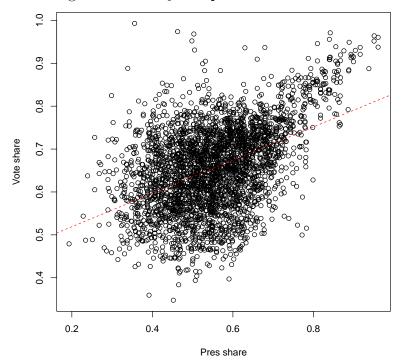


Figure 3: Scatter plot of presshare and voteshare.

Question 4

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 # (a) run regression where the outcome variable is \texttt{model_a.resid}
2 # and the explanatory variable is \texttt{model_b.resid}
3 model_d <- lm(model_a.resid ~ model_b.resid)</pre>
```

There is a positive and statistically reliable relationship between model_a.resid and model_b.resid, such that a one unit increase in the residual error from model A is associated with an average increase of 0.257 in residual error from model B (25.7%).

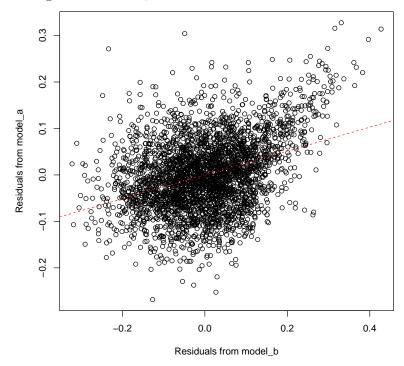
Table 4: Outcome variable is model_a.resid and the explanatory variable is model_b.resid.

	Model 1	
(Intercept)	-0.000	
	(0.001)	
$model_b.resid$	0.257^{***}	
	(0.012)	
\mathbb{R}^2	0.130	
$Adj. R^2$	0.130	
Num. obs.	3193	
RMSE	0.073	
*** $p < 0.001, **p < 0.01, *p < 0.05$		

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

```
plot(model_b.resid, model_a.resid,
ylab="Residuals from model_a", xlab="Residuals from model_b")
bline(model_d, col="red", lty=2)
```

Figure 4: Scatter plot of model_a.resid and model_b.resid.



3. Write the prediction equation.

$$model_a.resid = 0 + 0.257 \times model_b.resid$$

Question 5

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.

```
# (a) run regression where the outcome variable is \texttt{voteshare}

# and the explanatory variable is \texttt{difflog} and presvote

model_e <- lm(voteshare ~ difflog + presvote, data= inc.sub)
```

Table 5: Outcome variable is voteshare and the explanatory variables are difflog and presvote.

	Model 1	
(Intercept)	0.449***	
	(0.006)	
$\operatorname{difflog}$	0.036^{***}	
	(0.001)	
presvote	0.257^{***}	
	(0.012)	
\mathbb{R}^2	0.450	
$Adj. R^2$	0.449	
Num. obs.	3193	
RMSE	0.073	
***n < 0.001 **n < 0.01 *n < 0.05		

^{***}p < 0.001, **p < 0.01, *p < 0.05

2. Write the prediction equation.

$$voteshare = 0.449 + 0.036 \times difflog + 0.257 \times presvote$$

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

The coefficient for presvote in model_e is the same as the coefficient for model_b.resid in model_d because it represents the partial effect of presvote (the amount of covariation between presvote and voteshare that is unexplained by difflog).