A Statistical Analysis of Palm Beach County Election Votes in 2000

Case Study #1

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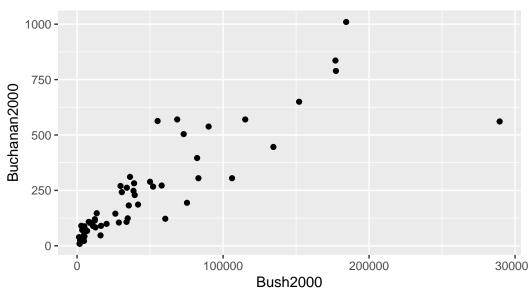
Introduction

During the U.S. presidential election of November 7, 2000, Democratic voters in Palm Beach County complained that a confusing butterfly ballot format caused them to accidentally vote for the Reform Party candidate Pat Buchanan instead of the Democratic candidate Gore. Additionally, 19,000 ballots were discarded because voters had marked two circles, possibly indicating that they got confused by the ballot and attempted to rectify their vote. Could the influx of votes for Buchanan in Palm Beach have occurred due to chance, or is there significant evidence that some external factor changed the outcome of the election? This analysis will examine Buchanan votes as a function of Bush votes in Florida counties in order to draw a conclusion about whether or not the Palm Beach county vote count could have happened by chance given the trend/distribution of Florida county votes.

Data Description

The election data consists of the numbers of votes for Buchanan and Bush in 67 counties in Florida. We exclude Palm Beach so that after we fit our model, we can see how the potentially abnormal Buchanan votes in that county compare to the rest of the data. The relationship between Bush and Buchanan votes is visualized in the scatter plot below

Association between votes for Bush and votes for Buchanan.



Relationship between Bush and Buchanan votes without Palm Beach county data

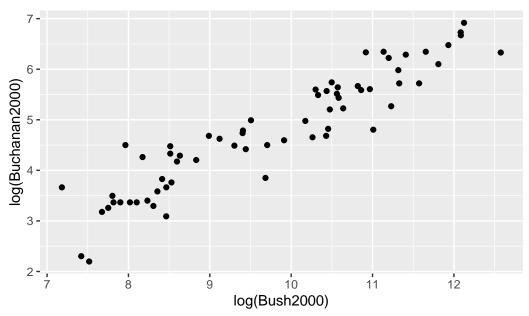
The average amount of Buchanan votes versus Bush votes is shown in the table below.

Average Buchanan Votes	Average Bush Votes	
210.76	41696.82	

Modeling Process

In order to create a model, we first observed a relationship in the scatter plot showing the votes for Buchanan and Bush in each of the Florida counties. From this, we noticed data clustering occurring in the bottom left corner of the scatter plot. We created diagnostic plots and checked to see if the model violated conditions for statistical inference. The diagnostic plots revealed violations of conditions of linearity, equal variance, and normality. In order to expand the scale and find a model that does not violate those conditions, we applied a log transformation to both the Buchanan and Bush votes, revealing a multiplicative relationship between the two. After transformation, the diagnostic plots reveal that regression conditions are met. Below is a plot of both variables transformed by log:





Let $buchanan_i$ denote the number of votes cast for Buchanan in county i and $bush_i$ denote the number of votes cast for Bush in Florida county i during the U.S. presidential election of November 7, 2000. Our final linear regression model for the mean is

$$E[log(buchanan_i)|log(bush_i)] = \beta_0 + \beta_1 \log(bush_i)$$

We fit our sample data to this model, and found estimates for the coefficients.

	Estimate	Std. Error	t value	$\Pr(> t)$
(Intercept)	-2.34	0.35	-6.61	0
$\log(\mathrm{Bush2000})$	0.73	0.04	20.32	0

Our sample intercept, $\hat{\beta}_0$, is -2.34149 with a standard error of 0.35442. Our sample slope, $\hat{\beta}_1$, is 0.73096 with a standard error of 0.03597. Both coefficients have a p-value less than 0.05, making them statistically significant. Thus, our fitted model is $\log(\widehat{buchanan}_i) = -2.34149 + 0.73096 \log(\widehat{bush}_i)$.

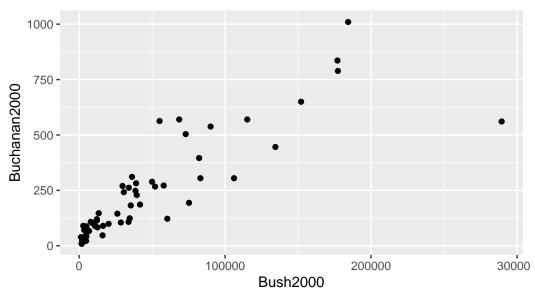
Following this, we created a 95% prediction interval for the number of predicted Buchanan votes in Palm Beach County during the 2000 election. Our prediction interval means that we are 95% confident that the number of votes cast for Buchanan in the U.S. presidential election of November 7, 2000 when the number of votes cast for Bush is 152,846 (the reported number of votes for Bush in Palm Beach county) is between 250 and 1399 votes. By computing the difference in votes between the prediction interval and the actual number of Buchanan votes, we determine that there are likely between 2008 to 3157 miscast ballots.

Conclusion

Since the 3407 votes for Buchanan in Palm Beach County in the 2000 election falls outside of our prediction interval of 250 to 1399 where Bush receives 152,846 votes, we can conclude that it is likely that there is some external factor impacting the votes of the residents in Palm Beach. One limitation of this analysis is the lack of information regarding the sociopolitical climate of Palm Beach County in 2020. We do not know the demographic breakdown of the county. If there were a higher percentage of centrists who would vote for a member of the Reform Party in 2020 then that could explain the abnormal amount of Buchanan votes. Additionally, we don't know what campaigning was like in Palm Beach County. There could have been additional campaigns for Buchanan there that caused a vote increase particularly in that county. Based on the assumption that some of the votes cast for Buchanan in Palm Beach were actually intended to be votes for Gore, we make a generalization that there are more Democratic voters in Palm Beach County than represented in our data. Without the cultural knowledge of the environment where the data comes from, we can't be fully confident that our model is a good fit for the data.

R Appendix

Association between votes for Bush and votes for Buchanan.



Relationship between Bush and Buchanan votes without Palm Beach county data

```
# Getting the average number of votes for each of the candidates excluding Palm

Beach
buchanan <- mean(election_wo_pb$Buchanan2000)
bush <- mean(election_wo_pb$Bush2000)

mean_election_wo_pb_table <- data.frame(Buchanan2000 = buchanan, Bush2000 = bush)

mean_election_wo_pb_table |> kbl(col.names = c("Mean Buchanan Votes", "Mean Bush

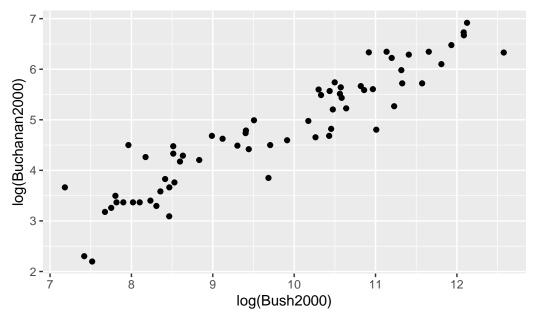
Votes"), align = "c", booktabs = T, linesep="", digits = c(2, 2, 2, 4)) |>

kable_classic(full_width = F, latex_options = c("HOLD_position"))
```

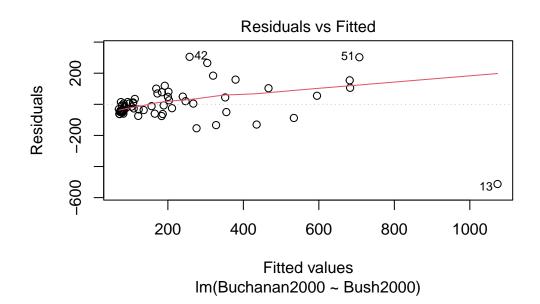
Mean Buchanan Votes	Mean Bush Votes	
210.76	41696.82	

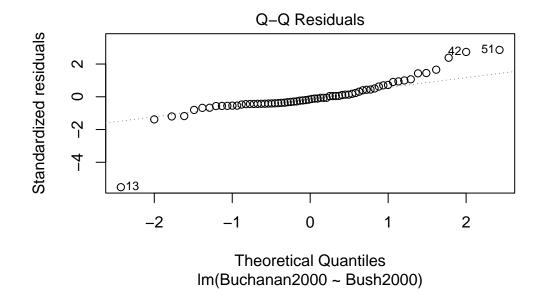
```
# Displaying a doubly transformed scatter plot between Bush and Buchanan votes election_wo_pb |> ggplot(aes(x = log(Bush2000), y = log(Buchanan2000))) + geom_point() + ggtitle("Transformed Association between votes for Bush and votes for Buchanan.")
```

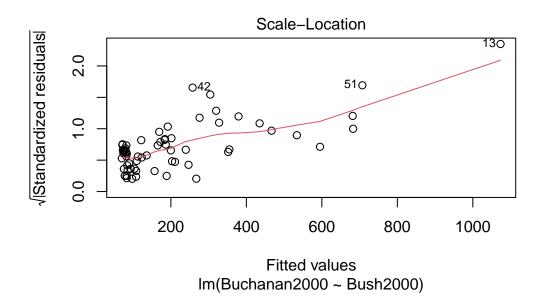


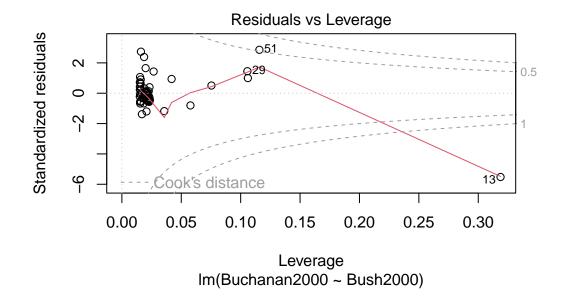


```
# Testing transformations to determine the best fit for the model
# Untransformed model
untransformed <- lm(Buchanan2000 ~ Bush2000, data = election_wo_pb)
plot(untransformed)</pre>
```

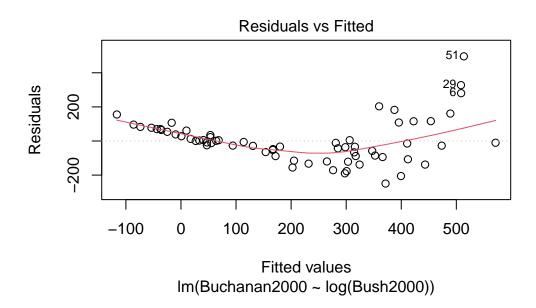


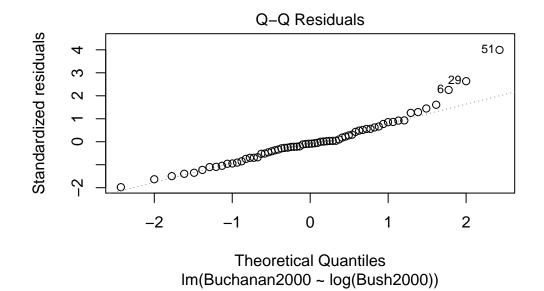


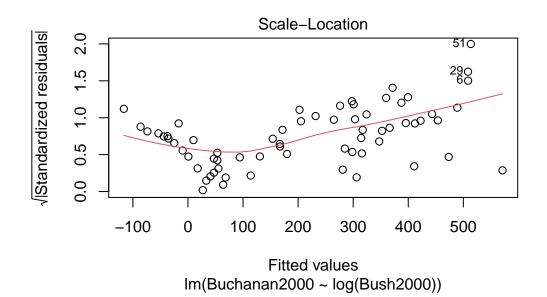


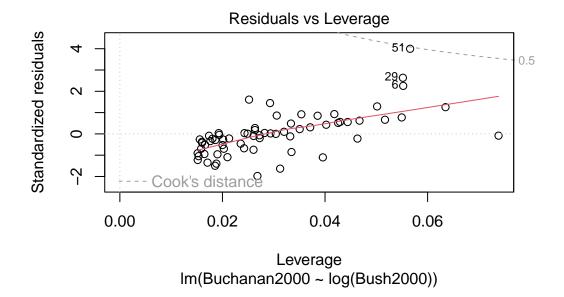


Model with a logarithmic explanatory variable transformation
xtransformed <- lm(Buchanan2000 ~ log(Bush2000), data = election_wo_pb)
plot(xtransformed)</pre>

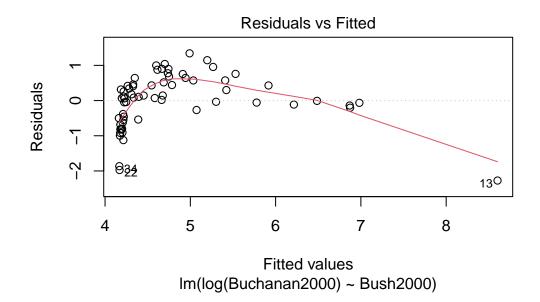


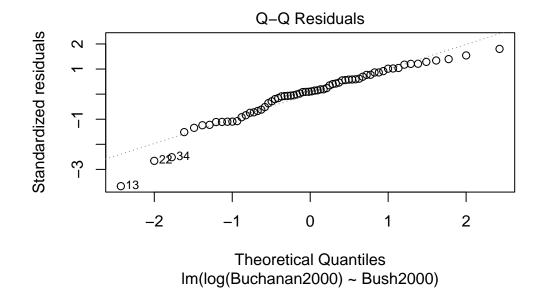


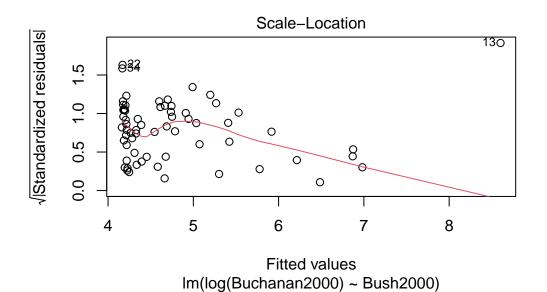


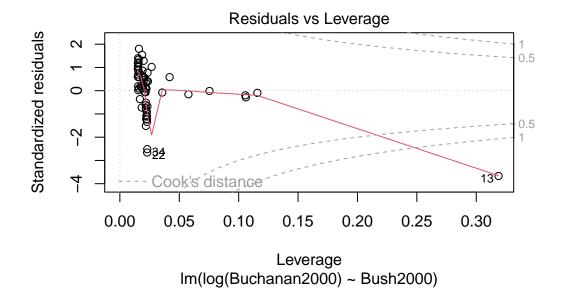


Model with a logarithmic response variable transformation
ytransformed <- lm(log(Buchanan2000) ~ Bush2000, data = election_wo_pb)
plot(ytransformed)</pre>

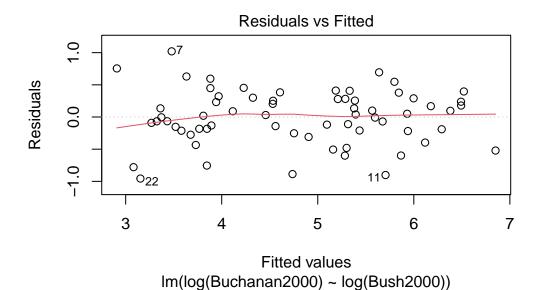


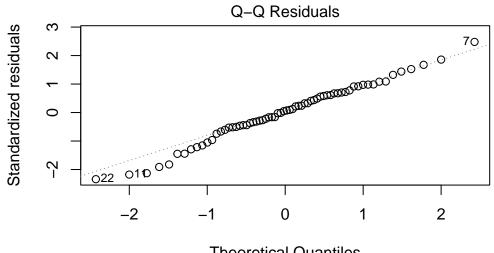




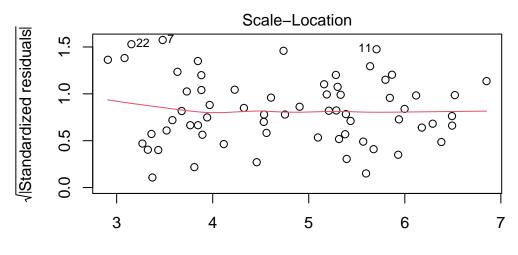


Model with a logarithmic explanatory and response variable transformation
both_transformed <- lm(log(Buchanan2000) ~ log(Bush2000), data = election_wo_pb)
plot(both_transformed)</pre>

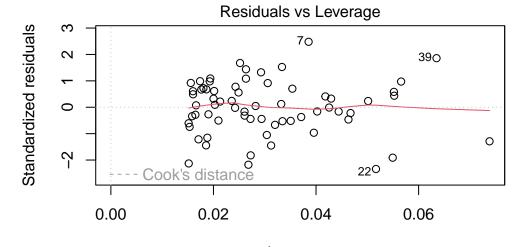




Theoretical Quantiles Im(log(Buchanan2000) ~ log(Bush2000))



Fitted values Im(log(Buchanan2000) ~ log(Bush2000))



Leverage Im(log(Buchanan2000) ~ log(Bush2000))

	Estimate	Std. Error	t value	$\Pr(> t)$
(Intercept)	-2.34	0.35	-6.61	0
$\log(\text{Bush2000})$	0.73	0.04	20.32	0

summary(both_transformed)

Call:

lm(formula = log(Buchanan2000) ~ log(Bush2000), data = election_wo_pb)

Residuals:

Min 1Q Median 3Q Max -0.95631 -0.21236 0.02503 0.28102 1.02056

Coefficients:

Estimate Std. Error t value Pr(>|t|)

```
(Intercept) -2.34149 0.35442 -6.607
                                              0.00000000907 ***
log(Bush2000) 0.73096 0.03597 20.323 < 0.000000000000000 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.4198 on 64 degrees of freedom
                            Adjusted R-squared: 0.8637
Multiple R-squared: 0.8658,
F-statistic: 413 on 1 and 64 DF, p-value: < 0.0000000000000000022
# Creating a prediction interval
predicted_palm_beach = data.frame(Bush2000 = 152846)
both_transformed |> augment(newdata = predicted_palm_beach, interval =

    "prediction", conf.level = 0.95)

# A tibble: 1 x 4
 Bush2000 .fitted .lower .upper
    <dbl> <dbl> <dbl> <dbl> <
1 152846 6.38 5.52 7.24
```