

hw12

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1.

d)

(a.)

$$\begin{aligned} \sum_{i=1}^n h_{ii} &= \text{tr}(H) = \text{tr}(X(X^T X)^{-1} X^T) \\ &= \text{tr}(X^T X (X^T X)^{-1}) \\ &= \text{tr}(I_p) \\ &= p \end{aligned}$$

cb) Since  $E(\hat{e}_i) = 0$ , so we have  $\hat{e}' \cdot \underline{1} = 0$  where  $\sum (\hat{e}_i) = 0$

$$\hat{e} = Y - \hat{Y} = Y - X(X^T X)^{-1} X^T Y = (I - H)Y \quad \hat{e}' \cdot \underline{1} = Y'(I - H) \cdot \underline{1}$$

$$\text{So, } (I - H) \cdot \underline{1} = 0 \quad \text{and} \quad H \cdot \underline{1} = \underline{1}$$

therefore,  $\underline{1} = \sum_{j=1}^n h_{ij}$ , Since  $H$  is symmetric,  $h_{ij} = h_{ji}$

$$\text{Thus, } \sum_{i=1}^n h_{ij} = \sum_{i=1}^n h_{ji} = 1$$

cc)

$$h_{ii} = \frac{1}{n} + (X_i^* - \bar{X})' (X^* X)^{-1} (X_i^* - \bar{X}) = \frac{1}{n} + \|A(X_i^* - \bar{X})\|^2, \text{ where } A = (X^* X)^{-1}$$

$$\text{So, } \|A(X_i^* - \bar{X})\|^2 > 0 \quad ; \quad \frac{1}{n} + \|A(X_i^* - \bar{X})\|^2 > \frac{1}{n} \quad \text{and} \quad h_{ii} > \frac{1}{n}$$

Since  $H$  is idempotent, \*

$$\begin{aligned} h_{ii} &= \sum_{j=1}^n h_{ij} h_{ji} = \sum_{j=1}^n h_{ij}^2 = \sum_{i \in J_i} h_{ij}^2 + h_{ii}^2 \quad J_i = \{j \in 1, \dots, n \mid x_i = x_j\} \\ &= \# \{J_i\} h_{ii}^2 + \sum_{j \in J_i} h_{ij}^2 \quad x_i = x_j \Rightarrow h_{ij} h_{ji} = h_{ii} \\ &= r h_{ii}^2 + \sum_{j \in J_i} h_{ij}^2 \quad r \text{ is the \# of rows} \\ &\geq r h_{ii}^2 \end{aligned}$$

$$\text{Thus } \frac{1}{n} \leq h_{ii} \leq \frac{1}{r} \quad \text{for } i = 1, \dots, n$$

## 2

```
library(alr4)
```

```
## Loading required package: car
```

```
## Loading required package: effects
```

```
##
```

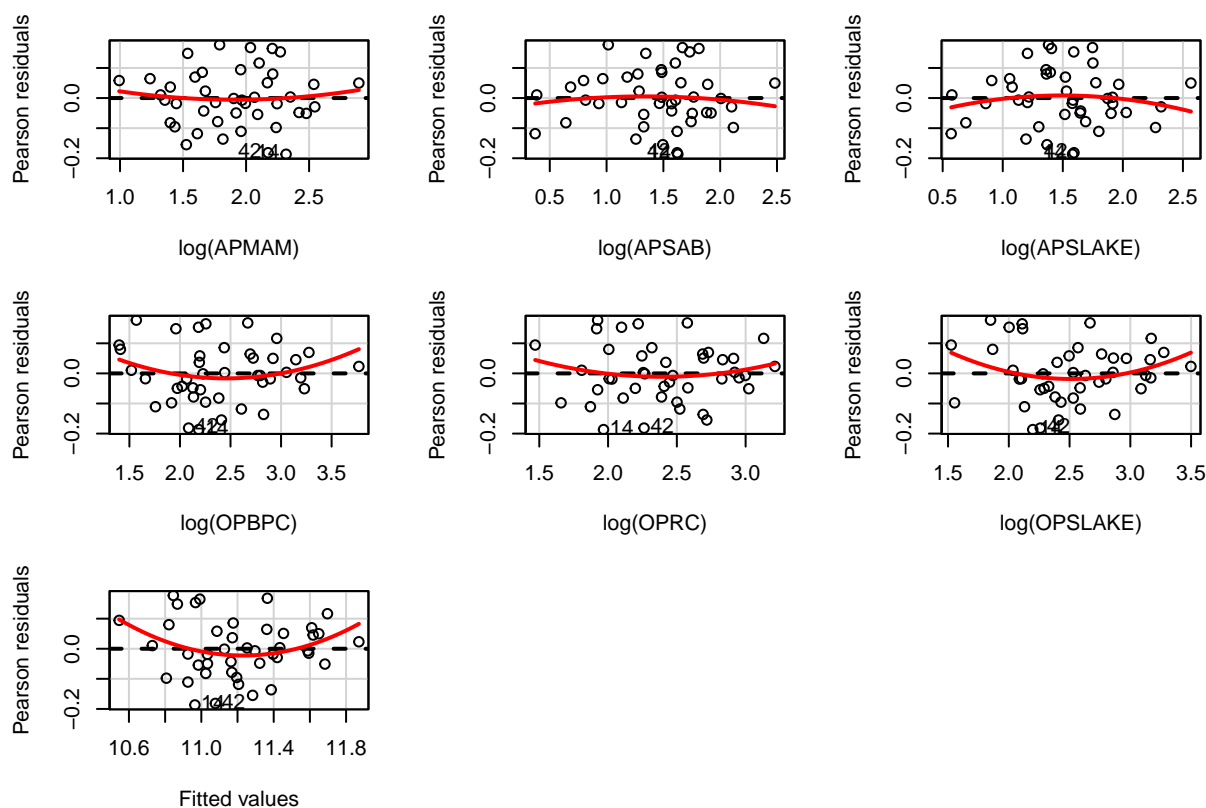
```
## Attaching package: 'effects'
```

```
## The following object is masked from 'package:car':
```

```
##
```

```
## Prestige
```

```
m1 = lm(formula = log(BSAAM)~log(APMAM)+log(APSAB)+log(APSLAKE)+log(OPBPC)+log(OPRC)+log(OPSLAKE), data=)
rp1=residualPlots(m1,id.n=2)
```



The residual plots shows that it seems to be a null plot.

```
rp1
```

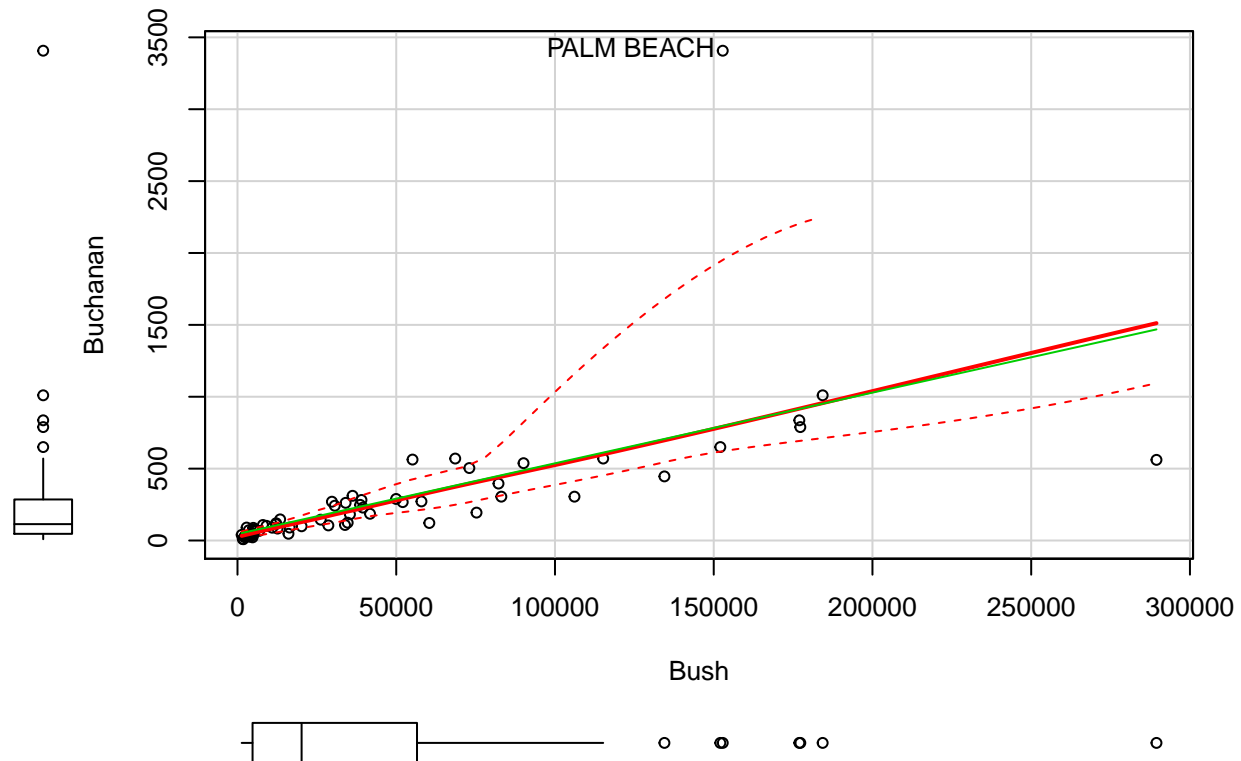
```
##          Test stat Pr(>|t|)
## log(APMAM)      0.450  0.656
## log(APSAB)     -0.465  0.645
## log(APSLAKE)   -0.852  0.400
## log(OPBPC)      1.385  0.175
## log(OPRC)       0.839  0.407
## log(OPSLAKE)    1.630  0.112
## Tukey test      1.839  0.066
```

None of the tests has small significance levels, providing no evidence against the mean function. We do not

have enough evidence to reject the  $H_0$  that there is no curvature.

### 3

```
m2=lm(Buchanan~Bush,data=florida)
scatterplot(Buchanan~Bush,data= florida,id.n=1)
```



```
## PALM BEACH
##          50
```

The Scatterplot shows **PALM BEACH** is an outlier.

```
outlierTest(m2)
```

```
##          rstudent unadjusted p-value Bonferonni p
## PALM BEACH 24.08014          8.6246e-34   5.7785e-32
```

```
cd2=cooks.distance(m2)
cd2[50]
```

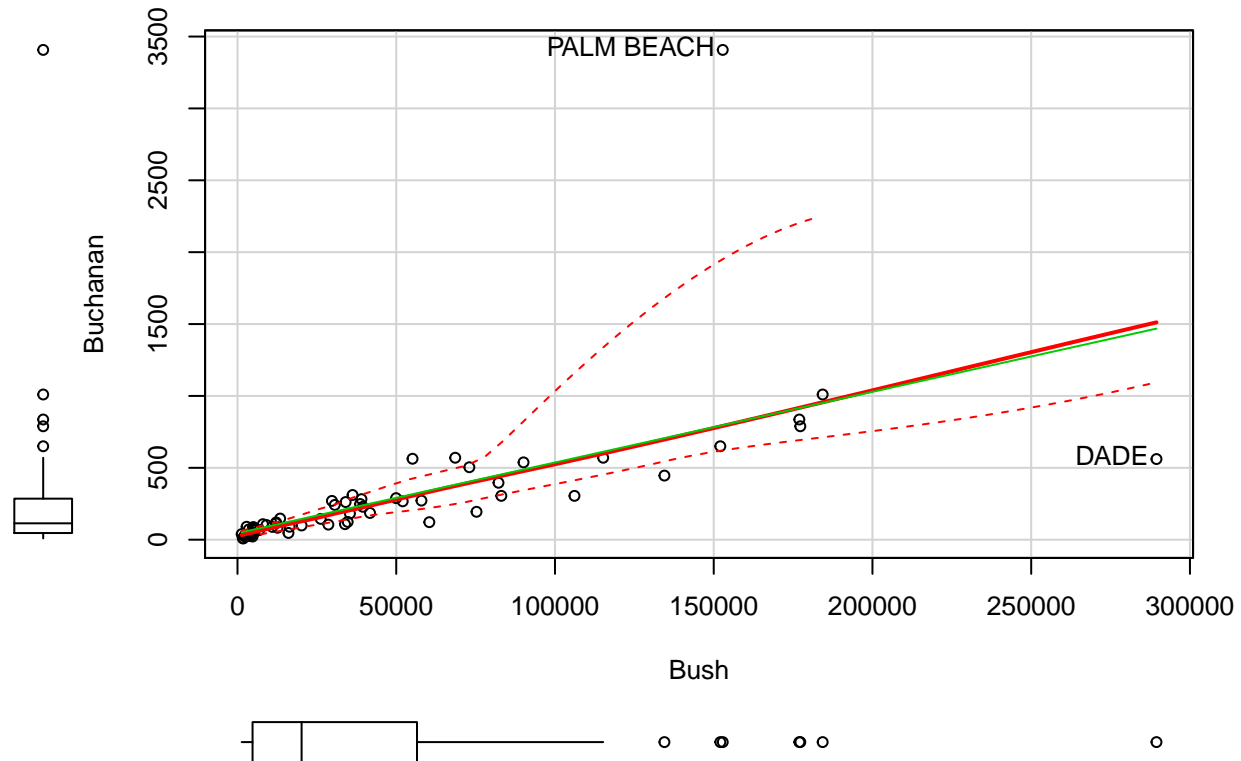
```
## PALM BEACH
##    2.231935
```

```
mean(cd2)
```

```
## [1] 0.06391973
```

Based on the test, we obtained a very small p-value that is an indication that it is an outlier. From the cook test for the city **PALM BEACH**, compare with the mean of **cd2** we observed a high value, and we conclude that the city **PALM BEACH** has a very high chance it is an outlier.

```
scatterplot(Buchanan~Bush,data= florida,id.n=2)
```



```
##      DADE PALM BEACH
##      13      50
```

It seems another country with an unusal value of the Buchanan vote, given its **Bush** value, is **DADE**

```
cd2[13]
```

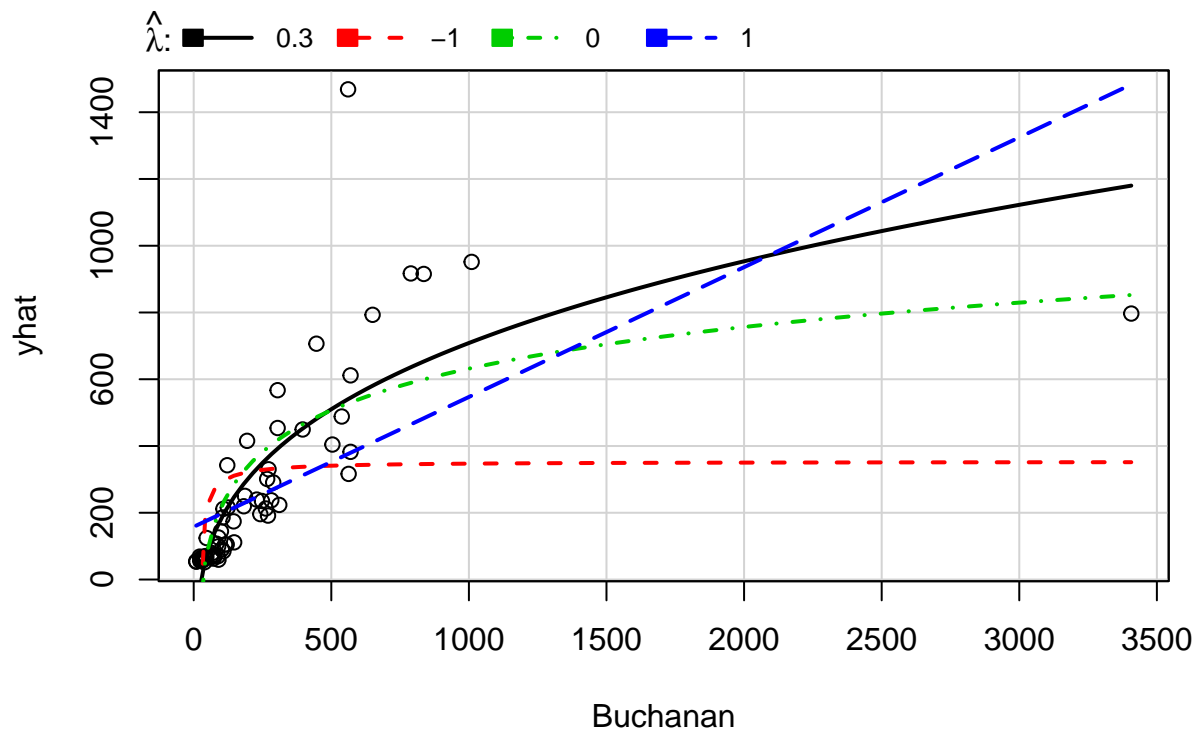
```
##      DADE
## 1.981366
```

```
mean(cd2)
```

```
## [1] 0.06391973
```

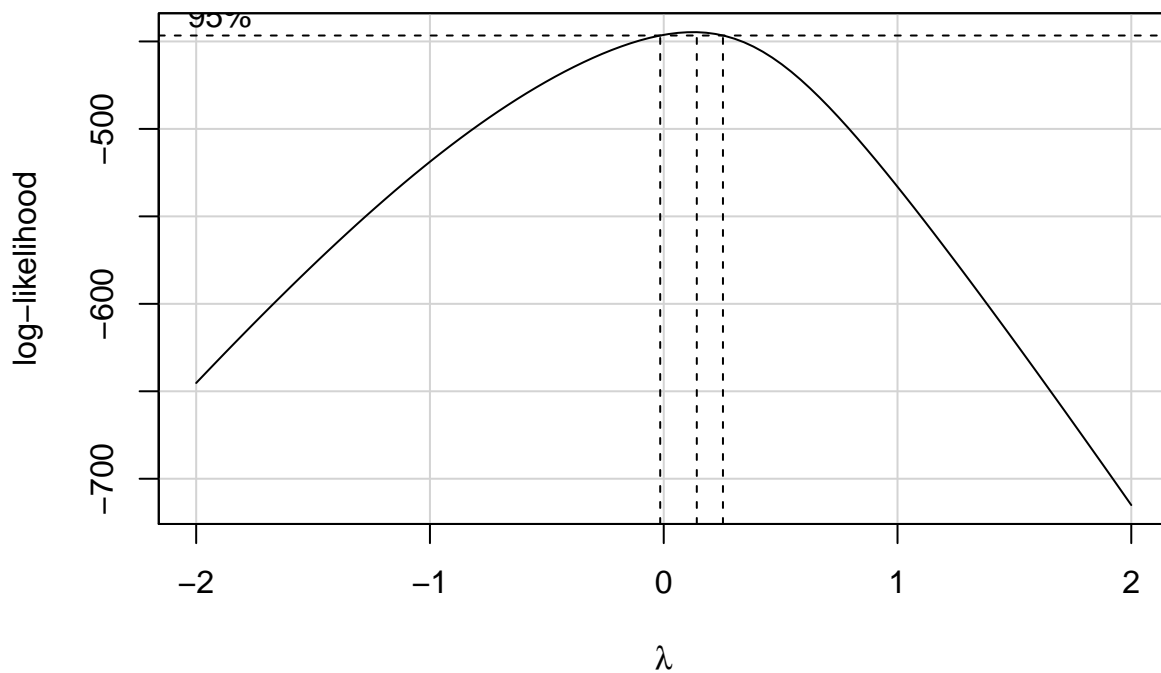
Same as the previous one, the distance value of the city **DADE** is large, we conclude that the city **DADE** has the high chance to be an outlier. It seems the butterfly ballot do have the issue of vote.

```
inverseResponsePlot(m2)
```



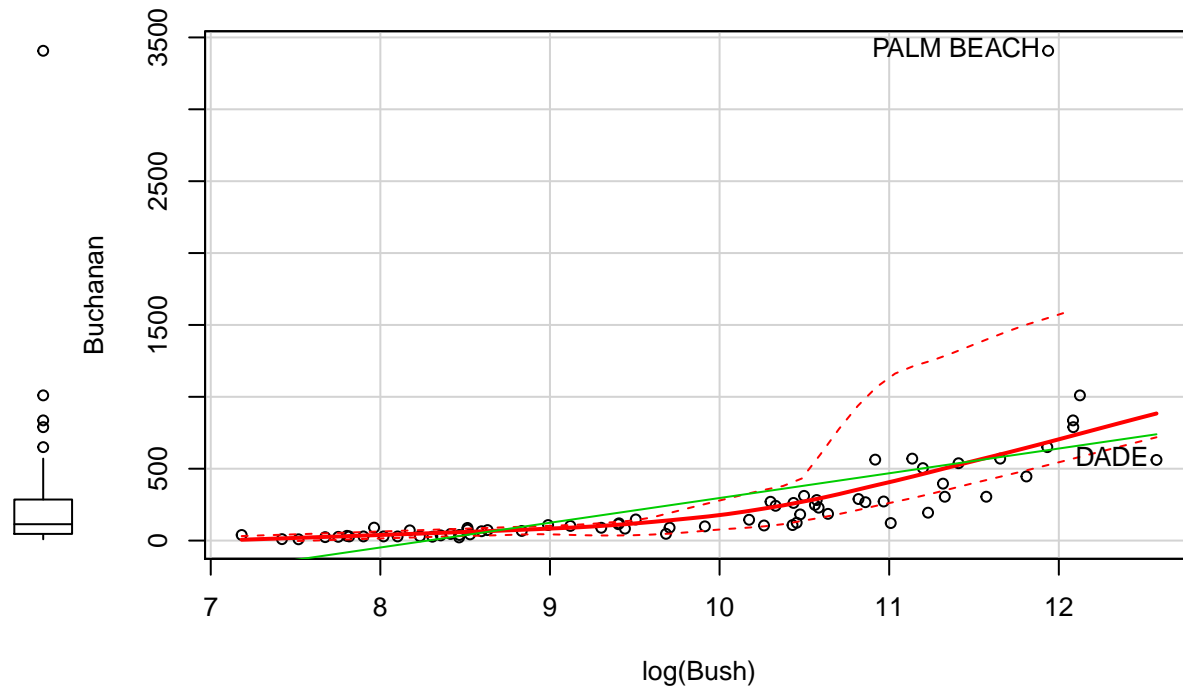
```
##      lambda      RSS
## 1  0.304032 1807862
## 2 -1.000000 4165308
## 3  0.000000 2099565
## 4  1.000000 3166621
```

```
boxCox(m2)
```



From the graph we observe that the best lambda is about 0.3. And the boxCox shows that 0 is concluded under the 95% confidence interval. Therefore, we use **log** transformation to better fit a simple linear regression.

```
m3=lm(Buchanan~log(Bush),data=florida)
scatterplot(Buchanan~log(Bush),data= florida,id.n=2)
```



```
##          DADE PALM BEACH
##          13          50
```

```
outlierTest(m3)
```

```
##          rstudent unadjusted p-value Bonferonni p
## PALM BEACH 22.25891          7.7958e-32    5.2232e-30
```

```
cd3=cooks.distance(m3)
cd3[50]
```

```
## PALM BEACH
##    1.448178
```

```
mean(cd3)
```

```
## [1] 0.0242152
```

Based on the cook distance test, and compared with mean value, the city **PALM BEACH** is still have large distance value and we still conclude that it has a high chance to be an outlier.

```
cd3[13]
```

```
##          DADE
## 0.009202589
```

```
mean(cd3)
```

```
## [1] 0.0242152
```

Comparing the previous steps to test the city **DADE**, We found that the DADE now has the distance below the mean, and we conclude that after the transformation, the city **DADE** seems no longer to be an outlier.