Fuel Example: Multiple Regression and Outliers

In this observational study (from Weisberg), we consider data for 48 states relating various state road and income variables (predictors) to per capital fuel consumption (response). Of particular interest is the relationship between fuel tax (predictor) and fuel consumption (response). After model fitting, one outlier (WY) is seen in the diagnostic plots.

Multiple Regression and Diagnostic Plots

```
library(dplyr)
library(car)
#In original file, State names are in first column.
#Using row.names will help identify states in the diagnostic plots below.
FuelData <- read.csv("~/Dropbox/STAT512/Lectures/MultReg1/MR1_Fuel.csv", row.names = 1)
str(FuelData)
## 'data.frame':
                    48 obs. of 8 variables:
##
   $ pop : int
                 1029 771 462 5787 968 3082 18366 7367 11926 10783 ...
                 9 9 9 7.5 8 10 8 8 8 7 ...
                 540 441 268 3060 527 1760 8278 4074 6312 5948 ...
  $ nlic : int
   $ inc
          : num
                 3.57 4.09 3.87 4.87 4.4 ...
   $ road : num 1.976 1.25 1.586 2.351 0.431 ...
  $ fuelc: int 557 404 259 2396 397 1408 6312 3439 5528 5375 ...
   $ dlic: num 52.5 57.2 58 52.9 54.4 57.1 45.1 55.3 52.9 55.2 ...
   $ fuel : int 541 524 561 414 410 457 344 467 464 498 ...
#Select columns of interest using select from dplyr.
FuelClean <- select(FuelData, fuel, tax, inc, road, dlic)</pre>
str(FuelClean)
  'data.frame':
                    48 obs. of 5 variables:
   $ fuel: int 541 524 561 414 410 457 344 467 464 498 ...
   $ tax : num
                9 9 9 7.5 8 10 8 8 8 7 ...
   $ inc : num 3.57 4.09 3.87 4.87 4.4 ...
                1.976 1.25 1.586 2.351 0.431 ...
   $ road: num
                52.5 57.2 58 52.9 54.4 57.1 45.1 55.3 52.9 55.2 ...
   $ dlic: num
cor(FuelClean)
##
                            tax
                                        inc
                                                   road
## fuel 1.00000000 -0.45128028 -0.24486207
                                            0.01904194
                                                         0.6989654
       -0.45128028 1.00000000
                                0.01266516 -0.52213014 -0.2880372
                                1.00000000
       -0.24486207 0.01266516
                                            0.05016279
## road 0.01904194 -0.52213014
                                0.05016279
                                            1.00000000 -0.0641295
## dlic 0.69896542 -0.28803717 0.15707008 -0.06412950 1.0000000
```

```
pairs(FuelClean)
                  5 6 7 8 9
                                                   5 10 15
                                                                               800
       fuel
                                                                               400
                                   tax
                                                                         മ
                       88
                            0
                         888
                                      inc
                                                    road
                                                                               9
   400 600 800
                                                                        65
                                3.0
                                      4.0
                                           5.0
                                                                   55
                                                               45
Model1 <- lm(fuel ~ tax, data = FuelData)</pre>
summary(Model1)
##
## Call:
## lm(formula = fuel ~ tax, data = FuelData)
##
## Residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
## -215.16 -72.27
                      6.74
                             41.28 355.74
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
                                   8.226 1.38e-10 ***
## (Intercept)
                 984.01
                          119.62
                            15.48 -3.430 0.00128 **
## tax
                 -53.11
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 100.9 on 46 degrees of freedom
## Multiple R-squared: 0.2037, Adjusted R-squared: 0.1863
## F-statistic: 11.76 on 1 and 46 DF, p-value: 0.001285
Model2 <- lm(fuel ~ tax + inc + road + dlic, data = FuelData)</pre>
summary(Model2)
##
## Call:
## lm(formula = fuel ~ tax + inc + road + dlic, data = FuelData)
```

##

```
## Residuals:
##
       Min
                  10
                      Median
                                     3Q
                                            Max
   -122.03
##
             -45.57
                      -10.66
                                 31.53
                                         234.95
##
##
   Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
##
                  377.291
                               185.541
                                          2.033 0.048207 *
##
  (Intercept)
                                         -2.682 0.010332 *
## tax
                  -34.790
                                12.970
##
  inc
                  -66.589
                                17.222
                                         -3.867 0.000368 ***
                   -2.426
                                         -0.716 0.477999
##
  road
                                 3.389
## dlic
                   13.364
                                 1.923
                                          6.950 1.52e-08 ***
##
                       '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
                     0
##
## Residual standard error: 66.31 on 43 degrees of freedom
## Multiple R-squared: 0.6787, Adjusted R-squared: 0.6488
## F-statistic: 22.71 on 4 and 43 DF, p-value: 3.907e-10
par(mfrow = c(2, 2))
plot(Model2)
                                                    Standardized residuals
                 Residuals vs Fitted
                                                                        Normal Q-Q
     200
                                      WYO
                                                                                             WYO
Residuals
                                                         3
                                       0
     -100
                400
                                                                 -2
                                                                               0
                                                                                      1
                                                                                             2
                       500
                              600
                                     700
                      Fitted values
                                                                      Theoretical Quantiles
Standardized residuals
                                                    Standardized residuals
                   Scale-Location
                                                                   Residuals vs Leverage
     2.0
                                                         က
                               OND
     1.0
                                                                                        NVO
                                                                          K's distance
     0.0
                                                             0.00
                                                                                            0.30
                400
                       500
                              600
                                     700
                                                                        0.10
                                                                                  0.20
                      Fitted values
                                                                           Leverage
```

Additional Diagnostics

The outlierTest() function from the car package runs an outlier test based on the Rstudent residual. Note that p-values are calculated with and without Bonferonni adjustment.

```
outlierTest(Model2)
```

rstudent unadjusted p-value Bonferonni p

```
## WY 4.490051
                      5.4704e-05
                                    0.0026258
Temp <- data.frame(State = row.names(FuelData),</pre>
                  Fuel = FuelData$fuel,
                  Pred = fitted(Model2),
                  ResidRaw = resid(Model2),
                  ResidStd = rstandard(Model2),
                  RStudent = rstudent(Model2))
#Reorder data by Rstudent using arrange from dplyr.
Temp <- arrange(Temp, desc(abs(RStudent)))</pre>
head(Temp)
##
    State Fuel
                   Pred
                         ResidRaw ResidStd RStudent
## 1
       WY 968 733.0528 234.94715 3.734462 4.490051
## 2
       RI 410 532.0289 -122.02893 -1.931710 -1.997765
## 3
       ND 714 596.4035 117.59655 1.842463 1.897347
## 4
       SD 865 772.9678 92.03225 1.542568 1.568543
## 5
       VA 547 460.2215 86.77850 1.359823 1.373781
## 6
       MA 414 493.3563 -79.35625 -1.279561 -1.289381
#Calculate Bonferonni adjusted p-value "By Hand".
\#Matched\ outlierTest\ output.
2*48*(1-pt(4.490, 48-4-2))
```

[1] 0.002626228