Lab3: Reducing Crime

w203 Lab3

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Exploratory Data Analysis

Perform an exploratory analysis to understand understand the determinants of crime and to generate policy suggestions that are applicable to local government.

Setup

First, we load the necessary libraries.

library(stargazer)

```
##
## Please cite as:
   Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary Statistics Tables.
   R package version 5.2.1. https://CRAN.R-project.org/package=stargazer
library(corrplot)
## corrplot 0.84 loaded
Then we load the dataset, which is in the same directory as this RMD.
# Load the data
rawCrimeData = read.csv("crime_v2.csv")
str(rawCrimeData)
  'data.frame':
                    97 obs. of 25 variables:
   $ county : int
                    1 3 5 7 9 11 13 15 17 19 ...
## $ year
                     87 87 87 87 87 87 87 87 87 87 ...
              : int
                     0.0356 0.0153 0.013 0.0268 0.0106 ...
   $ crmrte
              : num
                    0.298 0.132 0.444 0.365 0.518 ...
## $ prbarr : num
  $ prbconv : Factor w/ 92 levels "","\","0.068376102",..: 63 89 13 62 52 3 59 78 42 86 ...
##
   $ prbpris : num
                    0.436 0.45 0.6 0.435 0.443 ...
##
   $ avgsen
             : num
                     6.71 6.35 6.76 7.14 8.22 ...
##
   $ polpc
                     0.001828 0.000746 0.001234 0.00153 0.00086 ...
              : num
   $ density : num
                     2.423 1.046 0.413 0.492 0.547 ...
##
   $ taxpc
              : num
                     31 26.9 34.8 42.9 28.1 ...
##
   $ west
              : int
                     0 0 1 0 1 1 0 0 0 0 ...
##
   $ central : int
                     1 1 0 1 0 0 0 0 0 0 ...
                     0 0 0 0 0 0 0 0 0 0 ...
##
   $ urban
              : int
##
   $ pctmin80: num
                     20.22 7.92 3.16 47.92 1.8 ...
##
                     281 255 227 375 292 ...
   $ wcon
              : num
##
   $ wtuc
              : num
                     409 376 372 398 377 ...
##
  $ wtrd
              : num
                     221 196 229 191 207 ...
##
   $ wfir
                     453 259 306 281 289 ...
              : num
##
                     274 192 210 257 215 ...
   $ wser
              : num
                    335 300 238 282 291 ...
   $ wmfg
              : num
```

```
$ wfed
                     478 410 359 412 377 ...
              : num
##
                     292 363 332 328 367 ...
    $ wsta
              : num
                     312 301 281 299 343 ...
    $ wloc
              : num
                     0.0802 0.0302 0.4651 0.2736 0.0601 ...
##
   $ mix
              : num
    $ pctymle : num
                     0.0779 0.0826 0.0721 0.0735 0.0707 ...
```

The dataset contains 25 columns (variables) and 97 rows

Data Quality/Clean-up

```
apply(!is.na(rawCrimeData[,]), MARGIN = 2, mean)
                                                          prbpris
      county
                           crmrte
                                      prbarr
                                               prbconv
                                                                      avgsen
                   year
##
  0.9381443 0.9381443 0.9381443 1.0000000 0.9381443 0.9381443
       polpc
                density
##
                            taxpc
                                        west
                                                central
                                                            urban pctmin80
## 0.9381443 0.9381443 0.9381443 0.9381443 0.9381443 0.9381443 0.9381443
##
        wcon
                   wtuc
                             wtrd
                                        wfir
                                                   wser
                                                             wmfg
                                                                        wfed
## 0.9381443 0.9381443 0.9381443 0.9381443 0.9381443 0.9381443 0.9381443
##
                   wloc
                                     pctymle
        wsta
                              \min x
## 0.9381443 0.9381443 0.9381443 0.9381443
the dataset contains few NA that we'll need to fix before proceeding further.
complete.cases(rawCrimeData)
##
   [1]
         TRUE
               TRUE
                      TRUE
                            TRUE
                                  TRUE
                                         TRUE
                                               TRUE
                                                      TRUE
                                                            TRUE
                                                                  TRUE
                                                                         TRUE
                                                                  TRUE
                                                                         TRUE
## [12]
         TRUE
               TRUE
                      TRUE
                            TRUE
                                   TRUE
                                         TRUE
                                               TRUE
                                                      TRUE
                                                            TRUE
## [23]
         TRUE
               TRUE
                      TRUE
                            TRUE
                                   TRUE
                                         TRUE
                                               TRUE
                                                      TRUE
                                                            TRUE
                                                                  TRUE
                                                                         TRUE
## [34]
         TRUE
               TRUE
                      TRUE
                            TRUE
                                   TRUE
                                         TRUE
                                               TRUE
                                                      TRUE
                                                            TRUE
                                                                   TRUE
                                                                         TRUE
## [45]
         TRUE
               TRUE
                      TRUE
                            TRUE
                                   TRUE
                                         TRUE
                                               TRUE
                                                      TRUE
                                                            TRUE
                                                                  TRUE
                                                                         TRUE
               TRUE
  [56]
         TRUE
                      TRUE
                            TRUE
                                   TRUE
                                         TRUE
                                               TRUE
                                                      TRUE
                                                            TRUE
                                                                   TRUE
                                                                         TRUE
   [67]
         TRUE
               TRUE
                      TRUE
                            TRUE
                                   TRUE
                                         TRUE
                                               TRUE
                                                      TRUE
                                                            TRUE
                                                                  TRUE
                                                                         TRUE
##
   [78]
         TRUE
               TRUE
                      TRUE
                            TRUE
                                   TRUE
                                         TRUE TRUE TRUE
                                                            TRUE
                                                                   TRUE
                                                                         TRUE
## [89]
         TRUE
               TRUE
                      TRUE FALSE FALSE FALSE FALSE FALSE
crimeData <- rawCrimeData[complete.cases(rawCrimeData), ]</pre>
apply(!is.na(crimeData[,]), MARGIN = 2, mean)
##
     county
                        crmrte
                                  prbarr
                                          prbconv
                                                                         polpc
                 year
                                                   prbpris
                                                              avgsen
##
                                                 1
                    1
                             1
                                       1
                                                          1
                                                                    1
                                                                             1
##
                                 central
                                            urban pctmin80
    density
                taxpc
                          west
                                                                 wcon
                                                                          wtuc
##
          1
                    1
                             1
                                       1
                                                 1
                                                          1
                                                                    1
                                                                             1
##
       wtrd
                 wfir
                          wser
                                    wmfg
                                             wfed
                                                       wsta
                                                                 wloc
                                                                           mix
##
          1
                    1
                             1
                                       1
                                                 1
                                                          1
                                                                    1
                                                                             1
##
    pctymle
##
now we're good to go.
```

Univariate Variable Analysis of Key Variables

Crimes Committed

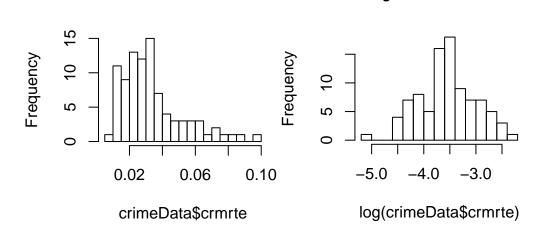
crmrte: crimes committed per person

Looking at the histogram of crime per person, the distribution appear to be positely skewed. Applying log() on crime shows the histogram to appear normally distributed.

```
hist(crimeData$crmrte, breaks=20, main = "Crimes Committed", cex.main=0.8)
hist(log(crimeData$crmrte), breaks=20, main = "Log Crimes Committed", cex.main=0.8)
```

Log Crimes Committed

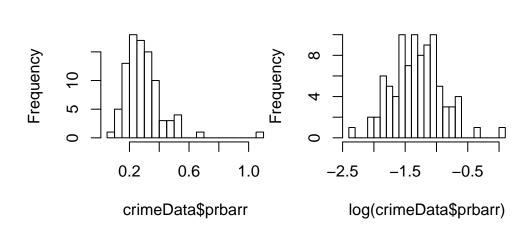
Log Probability of Arrest Log



prbarr: probability of arrest

Looking at the histogram of arrest per person, the distribution appear to be positely skewed. Applying log() shows the histogram to appear *less* normally distributed.

```
hist(crimeData$prbarr, breaks=20, main = "probability of Arrest", cex.main=0.8)
hist(log(crimeData$prbarr), breaks=20, main = "Log Probability of Arrest Log", cex.main=0.8)
```



prbconv: probability of conviction

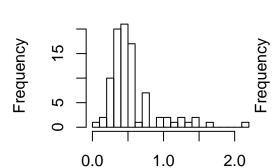
First, we'll need to concert the field from Factor to numeric for further analysis.

probability of Arrest

```
crimeData$prbconv <- as.numeric(as.character(crimeData$prbconv))</pre>
```

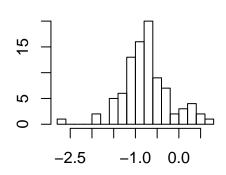
Looking at the histogram of probability of conviction, the distribution appear to be positely skewed. Applying log() shows the histogram to appear normally distributed.

```
crimeData$prbconv <- as.numeric(as.character(crimeData$prbconv))
hist(crimeData$prbconv, breaks=20, main = "Probability of Conviction", cex.main=0.8)
hist(log(crimeData$prbconv), breaks=20, main = "Log Probability of Conviction", cex.main=0.8)</pre>
```



Probability of Conviction

Log Probability of Conviction

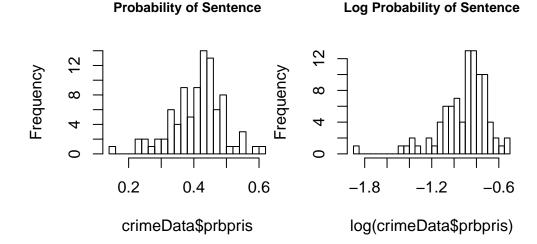


crimeData\$prbconv log(crimeData\$prbconv)

prbpris: of prison sentence

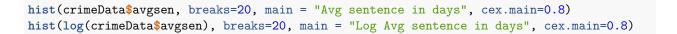
Looking at the histogram of probability of sentence, the distribution appear to be relatively normal. Applying log() shows the histogram to appear *less* normally distributed.

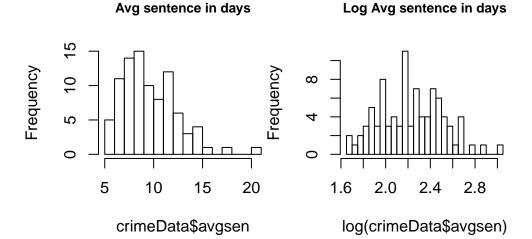
hist(crimeData\$prbpris, breaks=20, main = "Probability of Sentence", cex.main=0.8)
hist(log(crimeData\$prbpris), breaks=20, main = "Log Probability of Sentence", cex.main=0.8)



avgsen: avg. sentence, days

Looking at the histogram of probability of sentence, the distribution appear to be positively skewed. Applying log() shows the histogram to appear *more* normally distributed.

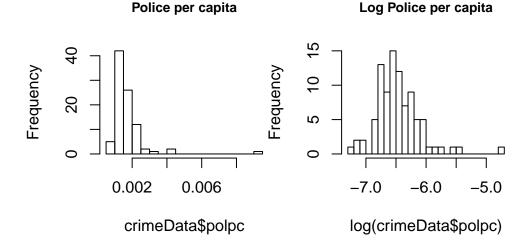




polpc: police per capita

Looking at the histogram of probability of sentence, the distribution appear to be positively skewed. Applying log() shows the histogram to appear *more* normally distributed.

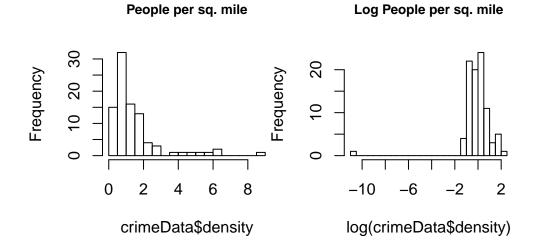
```
hist(crimeData$polpc, breaks=20, main = "Police per capita", cex.main=0.8)
hist(log(crimeData$polpc), breaks=20, main = "Log Police per capita", cex.main=0.8)
```



density: people per sq. mile

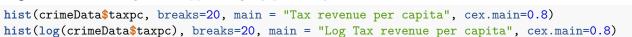
Looking at the histogram of probability of sentence, the distribution appear to be positevely skewed. Applying log() shows the histogram to appear *more* normally distributed, with one outlier.

```
hist(crimeData$density, breaks=20, main = "People per sq. mile", cex.main=0.8)
hist(log(crimeData$density), breaks=20, main = "Log People per sq. mile", cex.main=0.8)
```



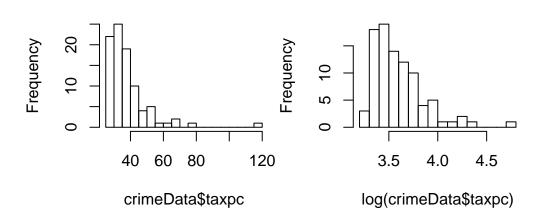
taxpc: tax revenue per capita

Looking at the histogram of probability of sentence, the distribution appear to be positevely skewed. Applying log() shows the histogram to appear slightly positevely skewed.



Log Tax revenue per capita

Tax revenue per capita

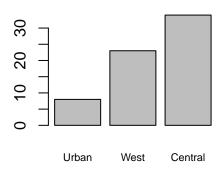


taxpc: tax revenue per capita

Looking at the histogram of probability of sentence, the distribution appear to be positevely skewed. Applying log() shows the histogram to appear slightly positevely skewed.

```
barplot(c(sum(crimeData$urban), sum(crimeData$west), sum(crimeData$central)),
  names.arg = c("Urban", "West", "Central"), main = "Part of the state counties are in", cex.main=0.8,
```

Part of the state counties are in

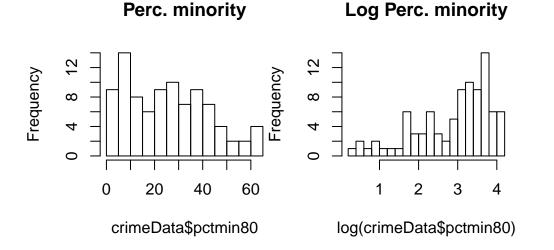


pctmin80: perc. minority, 1980

Looking at the histogram of probability of sentence, the distribution appear to be slightly positively skewed. Applying log() shows the histogram to appear negatively skewed.

```
hist(crimeData$pctmin80, breaks=20, main = "Perc. minority")
hist(log(crimeData$pctmin80), breaks=20, main = "Log Perc. minority")
```

Perc. minority



wcon: weekly wage, construction

Looking at the histogram of probability of sentence, the distribution appear to be slightly positively skewed. Applying log() shows the histogram to appear normally distributed

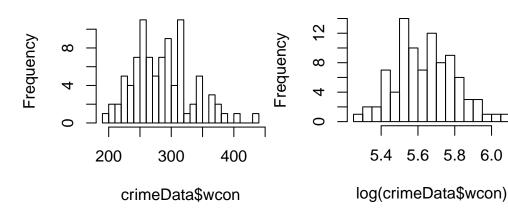
```
hist(crimeData$wcon, breaks=20, main = "Weekly wage (Construction)", cex.main=.8)
hist(log(crimeData$wcon), breaks=20, main = "Log Weekly wage (Construction)", cex.main=.8)
```

Weekly wage (Construction)

Log Weekly wage (Construction)

5.8

6.0

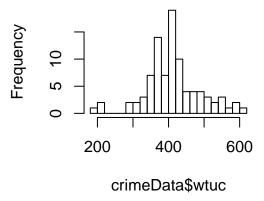


wtuc: wkly wge, trns, util, commun

Looking at the histogram, the distribution appear to be normally distributed.

hist(crimeData\$wtuc, breaks=20, main = "Weekly wage (trns/util/commun)", cex.main=.8)

Weekly wage (trns/util/commun)



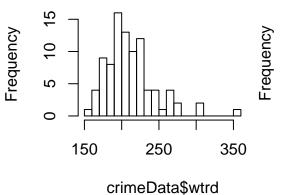
wtrd: wkly wge, whlesle, retail trade

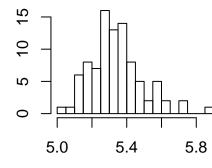
Looking at the histogram, the distribution appear to be positively skewed. Applying log() shows the histogram to appear normally distributed

hist(crimeData\$wtrd, breaks=20, main = "Weekly wage (whlesle, retail)", cex.main=.8) hist(log(crimeData\$wtrd), breaks=20, main = "Log Weekly wage (whlesle, retail)", cex.main=.8)

Weekly wage (whlesle, retail)

Log Weekly wage (whiesle, retail)





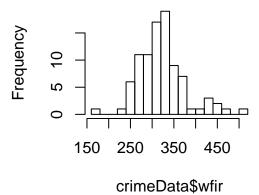
log(crimeData\$wtrd)

wfir: wkly wge, fin, ins, real est

Looking at the histogram, the distribution appear to be normally distributed.

hist(crimeData\$wfir, breaks=20, main = "Weekly wage (wge, fin, ins, real est)", cex.main=.8)

Weekly wage (wge, fin, ins, real est)



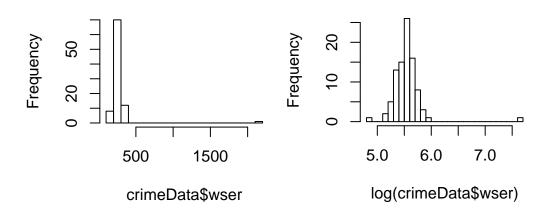
wser: wkly wge, service industry

Looking at the histogram, the distribution appear to be positively skewed. Applying log() shows the histogram to appear normally distributed with one outlier.

hist(crimeData\$wser, breaks=20, main = "Weekly wage (service industry)", cex.main=.8)
hist(log(crimeData\$wser), breaks=20, main = "Log Weekly wage (service industry)", cex.main=.8)



Log Weekly wage (service industry)



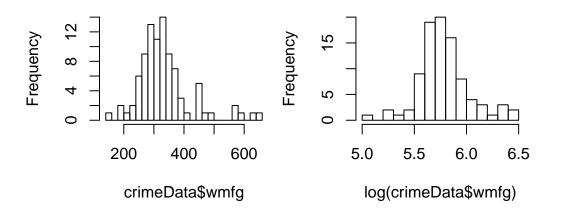
wmfg: wkly wge, manufacturing

Looking at the histogram, the distribution appear to be slighly positively skewed. Applying log() shows the histogram to appear normally distributed.

hist(crimeData\$wmfg, breaks=20, main = "Weekly wage (manufacturing)", cex.main=.8)
hist(log(crimeData\$wmfg), breaks=20, main = "Log Weekly wage (manufacturing)", cex.main=.8)

Weekly wage (manufacturing)

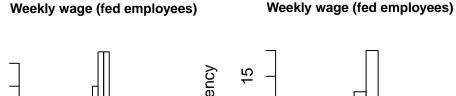
Log Weekly wage (manufacturing)

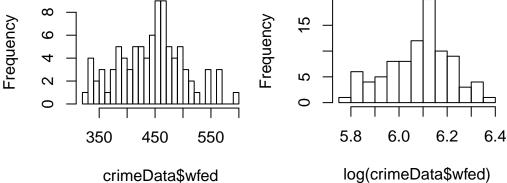


wfed: wkly wge, fed employees

Applying log() shows the histogram to appear normally distributed.

```
hist(crimeData$wfed, breaks=20, main = "Weekly wage (fed employees)", cex.main=.8)
hist(log(crimeData$wfed), breaks=20, main = "Weekly wage (fed employees)", cex.main=.8)
```

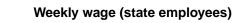




wsta: wkly wge, state employees

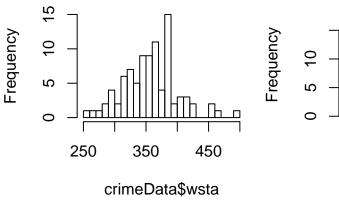
Looking at the histogram, the distribution appear to be slighly positively skewed. Applying log() shows the histogram to appear normally distributed.

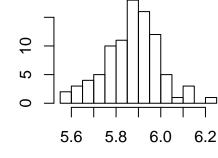
hist(crimeData\$wsta, breaks=20, main = "Weekly wage (state employees)", cex.main=.8) hist(log(crimeData\$wsta), breaks=20, main = "Log Weekly wage (state employees)", cex.main=.8)



Log Weekly wage (state employees)

log(crimeData\$wsta)



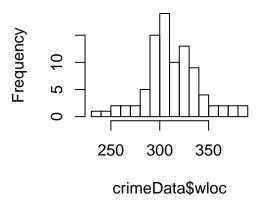


wloc: wkly wge, local gov emps

Looking at the histogram, the distribution appear to be slighly normally distributed.

hist(crimeData\$wloc, breaks=20, main = "Weekly wage (local gov employees)", cex.main=.8)

Weekly wage (local gov employees)



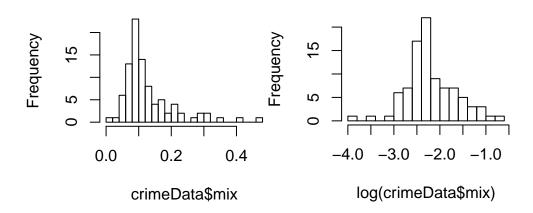
mix: offense mix: face-to-face/other

Looking at the histogram, the distribution appear to be slighly positively skewed. Applying log() shows the histogram to appear normally distributed.

```
hist(crimeData$mix, breaks=20, main = "Face-to-face/other", cex.main=.8)
hist(log(crimeData$mix), breaks=20, main = "Face-to-face/other", cex.main=.8)
```

Face-to-face/other

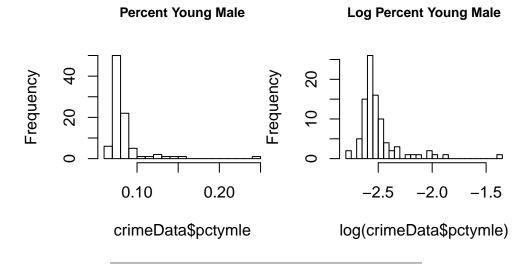
Face-to-face/other



pctymle: percent young male

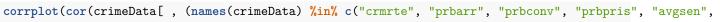
Looking at the histogram, the distribution appear to be positively skewed. Applying log() shows the histogram to appear positively skewed with one outlier.

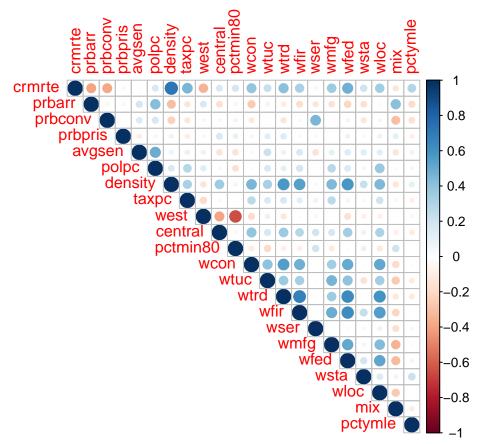
```
hist(crimeData$pctymle, breaks=20, main = "Percent Young Male", cex.main=.8)
hist(log(crimeData$pctymle), breaks=20, main = "Log Percent Young Male", cex.main=.8)
```



Analysis of Key Relationships

It is very imperative to realize the relationship between crime rate and all the data available to us. We'll use corrplot to make the exploration of key relationships clearer.





The above plot indicates the following *positive* relationships with crime rate:

- 1. Density (density).
- 2. Tax revenue per capita (taxpc).
- 3. All wage varibles.
- 4. Young Male (pctymle)

The above plot also indicates the following *negative* relationships with crime rate:

- 1. Probability of Arrest (prbarr)
- 2. Probablity of Conviction (prbconv)
- 3. West region of NC (west)

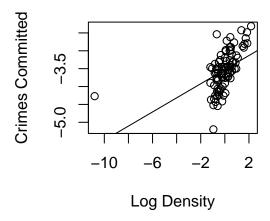
Crimes Committed per person (crmrte) & People per sq. (density)

As you can see from the correlation plot below, there is a positive linear relationship between crime rate and density.

```
plot(log(crimeData$density), log(crimeData$crmrte),
    main="Crime Density vs Crime Rate",
    xlab="Log Density",
    ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$density)))
cor(crimeData$crmrte, crimeData$density)
```

[1] 0.7289632

Crime Density vs Crime Rate

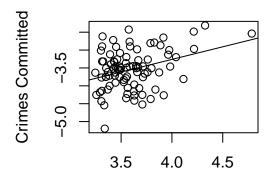


Crimes Committed per person (crmrte) & Tax revenue per capita (taxpc)

```
plot(log(crimeData$taxpc), log(crimeData$crmrte),
    main="Tax revenue per capita vs Crime Rate",
    xlab="Tax revenue per capita",
    ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$taxpc)))
cor(crimeData$crmrte, crimeData$taxpc)
```

[1] 0.4509798

Tax revenue per capita vs Crime Rate



Tax revenue per capita

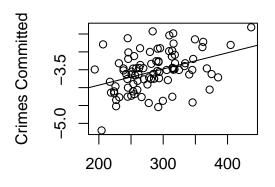
Crimes Committed per person (crmrte) & Wages

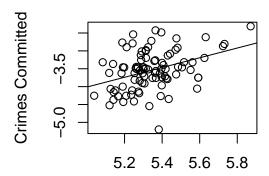
```
plot(crimeData$wcon, log(crimeData$crmrte),
   main="Wekly Wages (trns, util, commun) vs Crime Rate",
   xlab="Wekly Wages",
   ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ crimeData$wcon))
plot(log(crimeData$wtrd), log(crimeData$crmrte),
    main="Wekly Wages (whlesle, retail trade) vs Crime Rate",
   xlab="Wekly Wages",
   ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$wtrd)))
plot(crimeData$wfir, log(crimeData$crmrte),
    main="Wekly Wages (fin, ins, real est) vs Crime Rate",
    xlab="Wekly Wages",
    ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ crimeData$wfir))
plot(log(crimeData$wser), log(crimeData$crmrte),
    main="Wekly Wages (service industry) vs Crime Rate",
    xlab="Wekly Wages",
   ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$wser)))
plot(log(crimeData$wmfg), log(crimeData$crmrte),
   main="Wekly Wages (manufacturing) vs Crime Rate",
   xlab="Wekly Wages",
    ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$wmfg)))
plot(log(crimeData$wfed), log(crimeData$crmrte),
   main="Wekly Wages (fed employees) vs Crime Rate",
   xlab="Wekly Wages",
   ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$wfed)))
```

```
plot(crimeData$wsta, log(crimeData$crmrte),
    main="Wekly Wages (state employees) vs Crime Rate",
    xlab="Wekly Wages",
    ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ crimeData$wsta))

plot(log(crimeData$wloc), log(crimeData$crmrte),
    main="Wekly Wages (local gov emps) vs Crime Rate",
    xlab="Wekly Wages",
    ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$wloc)))
```

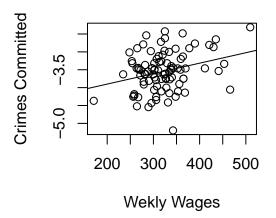
Wekly Wages (trns, util, commun) vs Crime | Wekly Wages (whlesle, retail trade) vs Crime

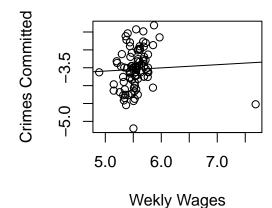




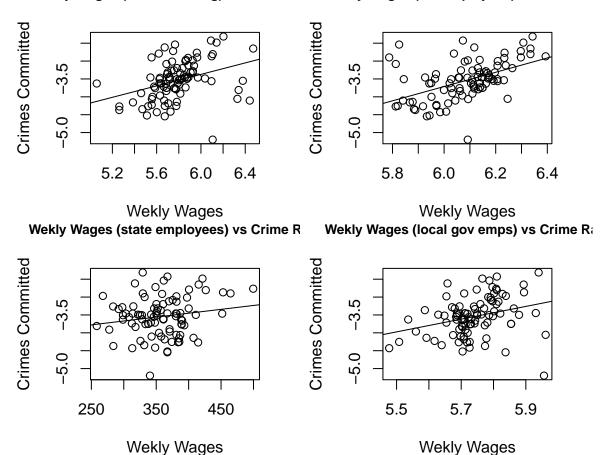
Wekly Wages
Wekly Wages (fin, ins, real est) vs Crime Ra

Wekly Wages
Wekly Wages (service industry) vs Crime R





Wekly Wages (manufacturing) vs Crime Ra Wekly Wages (fed employees) vs Crime Ra

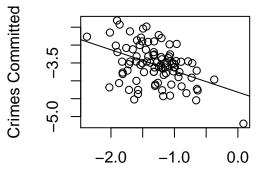


Crimes Committed per person (crmrte) & Probabiliy of Arrest (prbarr)

```
plot(log(crimeData$prbarr), log(crimeData$crmrte),
    main="Probability of Arrest vs Crime Rate",
    xlab="Probability of Arrest",
    ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$prbarr)))
cor(crimeData$crmrte, crimeData$prbarr)
```

[1] -0.3933297

Probabiliy of Arrest vs Crime Rate



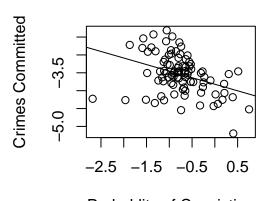
Probabiliy of Arrest

Crimes Committed per person (crmrte) & Tax revenue per capita (prbconv)

```
plot(log(crimeData$prbconv), log(crimeData$crmrte),
    main="Probablity of Conviction vs Crime Rate",
    xlab="Probablity of Conviction",
    ylab="Crimes Committed", cex.main=0.8)
abline(lm(log(crimeData$crmrte) ~ log(crimeData$prbconv)))
cor(crimeData$crmrte, crimeData$prbconv)
```

[1] -0.3859724

Probablity of Conviction vs Crime Rate



Probablity of Conviction

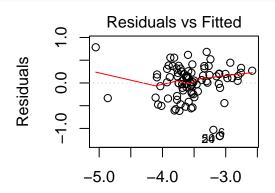
Proposed Models

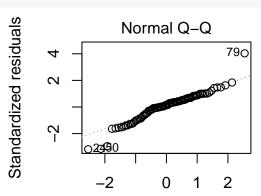
Model 1: with only the explanatory variables

Using a combination of key positive and negative attributes to crime rate, we're recommending the following model:

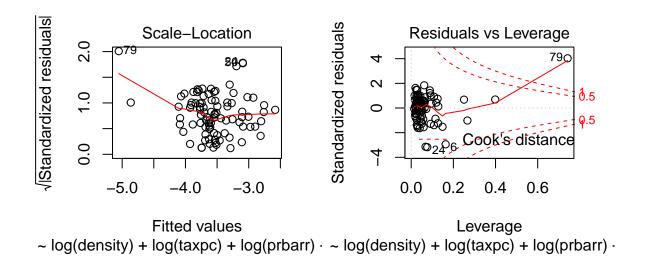
```
crimeDeterm = \beta_0 + \beta_1 \cdot log(density) + \beta_2 \cdot log(taxpc) + \beta_3 \cdot log(prbarr) + \beta_4 \cdot log(prbconv) + \beta_5 \cdot log(pctymle)
```

```
model1 <- lm(log(crmrte) ~ log(density) + log(taxpc) + log(prbarr) + log(prbconv)</pre>
             + log(pctymle), data=crimeData)
summary(model1)
##
## Call:
   lm(formula = log(crmrte) ~ log(density) + log(taxpc) + log(prbarr) +
##
       log(prbconv) + log(pctymle), data = crimeData)
##
  Residuals:
##
##
        Min
                  1Q
                       Median
                                        0.78454
  -1.17009 -0.18998 0.05217 0.22241
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -5.27049
                            0.77309
                                    -6.817 1.27e-09 ***
## log(density) 0.12180
                            0.03208
                                      3.797 0.000274 ***
## log(taxpc)
                 0.41512
                            0.16111
                                      2.577 0.011705 *
## log(prbarr) -0.47872
                            0.11901
                                    -4.023 0.000124 ***
## log(prbconv) -0.35278
                            0.08166
                                     -4.320 4.21e-05 ***
                                      1.135 0.259713
## log(pctymle) 0.25870
                            0.22800
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.383 on 85 degrees of freedom
## Multiple R-squared: 0.5354, Adjusted R-squared: 0.5081
## F-statistic: 19.59 on 5 and 85 DF, p-value: 6.212e-13
plot(model1)
```





Fitted values Theoretical Quantiles ~ log(density) + log(taxpc) + log(prbarr) · ~ log(density) + log(taxpc) + log(prbarr) ·

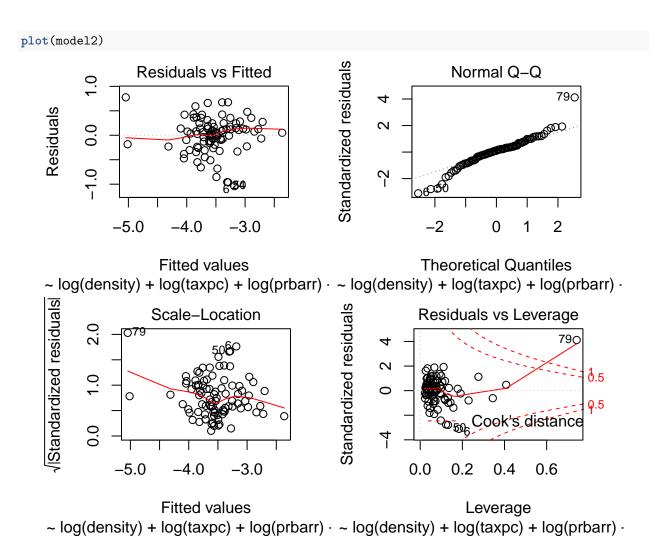


Model 2: with key explanatory variables and only covariates

In this model, we'll include the variables (avgsen, mix), as we think they will contribute to the accuracy of your results without introducing substantial bias.

```
crimeDeterm = \beta_0 + \beta_1 \cdot log(density) + \beta_2 \cdot log(taxpc) + \beta_3 \cdot log(prbarr) + \beta_4 \cdot log(prbconv) + \beta_5 \cdot log(pctymle) + \beta_6 \cot log(avgsen) + \beta_
```

```
model2 <- lm(log(crmrte) ~ log(density) + log(taxpc) + log(prbarr) + log(prbconv)</pre>
             + log(pctymle) + log(avgsen) + log(mix), data=crimeData)
summary(model2)
##
## Call:
  lm(formula = log(crmrte) ~ log(density) + log(taxpc) + log(prbarr) +
       log(prbconv) + log(pctymle) + log(avgsen) + log(mix), data = crimeData)
##
##
## Residuals:
##
                  1Q
                       Median
                                    3Q
  -1.04133 -0.17272 0.03557
                              0.18088
                                        0.77601
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                -4.57642
                            0.84429
                                     -5.420 5.71e-07 ***
## log(density)
                 0.14290
                            0.03216
                                      4.443 2.72e-05 ***
## log(taxpc)
                 0.43437
                            0.15707
                                      2.765 0.00700 **
## log(prbarr)
                -0.59111
                            0.12448
                                     -4.748 8.47e-06 ***
## log(prbconv) -0.26492
                            0.08512
                                     -3.112
                                             0.00255 **
                                      1.494
## log(pctymle)
                 0.33420
                            0.22370
                                             0.13898
## log(avgsen)
                -0.04673
                            0.14307
                                     -0.327
                                             0.74477
## log(mix)
                 0.24922
                            0.09205
                                      2.708
                                             0.00823 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3707 on 83 degrees of freedom
## Multiple R-squared: 0.575, Adjusted R-squared: 0.5392
## F-statistic: 16.04 on 7 and 83 DF, p-value: 3.657e-13
```

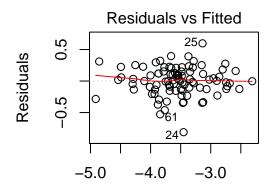


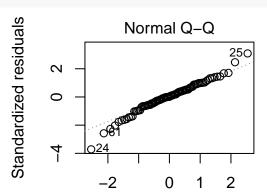
Model 3: includes the previous covariates, and most, if not all, other covariates

In this model, we'll include all the data available to us to demonstrate the robustness of of results to model specification.

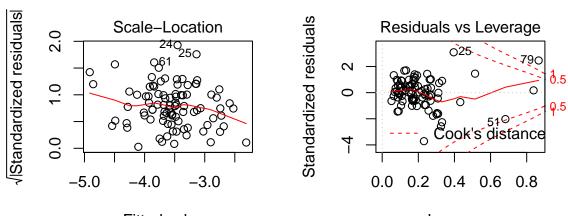
 $crimeDeterm = \beta_0 + \beta_1 \cdot log(density) + \beta_2 \cdot log(taxpc) + \beta_3 \cdot log(prbarr) + \beta_4 \cdot log(prbconv) + \beta_5 \cdot log(pctymle) + \beta_6 \cdot log(avgsen) + \beta_7 \cdot log(prbarr) + \beta_8 \cdot log(prbarr) + \beta_$

```
##
       log(wser) + log(wmfg) + log(wfed) + log(wsta) + wloc, data = crimeData)
##
##
  Residuals:
##
        Min
                                     3Q
                  1Q
                       Median
                                             Max
##
   -0.81225 -0.12715
                      0.00293
                                0.14931
##
  Coefficients:
##
##
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 -7.074e+00
                             2.935e+00
                                         -2.410 0.01852 *
  log(density)
                  1.306e-01
                             3.127e-02
                                          4.175 8.24e-05 ***
## log(taxpc)
                  5.837e-02
                             1.361e-01
                                          0.429
                                                 0.66929
## log(prbarr)
                 -5.376e-01
                             8.715e-02
                                         -6.168 3.62e-08 ***
## log(prbconv)
                 -3.022e-01
                             6.668e-02
                                         -4.532 2.27e-05 ***
                  1.612e-01
## log(pctymle)
                             1.703e-01
                                          0.946
                                                0.34708
## log(avgsen)
                 -2.993e-01
                             1.165e-01
                                         -2.570
                                                 0.01224 *
## log(mix)
                  6.474e-02
                             7.158e-02
                                          0.904
                                                 0.36879
## log(prbpris)
                 -3.134e-01
                             1.444e-01
                                         -2.171
                                                0.03326 *
## log(polpc)
                  4.777e-01
                             1.126e-01
                                          4.241 6.51e-05
                  2.278e-01
                             3.330e-02
                                          6.841 2.17e-09 ***
## log(pctmin80)
## log(wcon)
                  2.293e-01
                             2.222e-01
                                          1.032
                                                0.30568
## log(wtrd)
                  4.131e-01
                             3.036e-01
                                          1.361
                                                 0.17781
## wfir
                 -1.347e-03
                             7.957e-04
                                         -1.693
                                                 0.09469
## log(wser)
                                         -2.801
                 -3.086e-01
                             1.102e-01
                                                 0.00654 **
                  8.056e-02
                                          0.520
## log(wmfg)
                             1.548e-01
                                                 0.60436
## log(wfed)
                  6.546e-01
                             3.402e-01
                                          1.924
                                                 0.05827
## log(wsta)
                 -6.868e-03
                             2.660e-01
                                         -0.026
                                                 0.97947
## wloc
                 -2.877e-05
                             1.420e-03
                                         -0.020
                                                 0.98389
##
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2493 on 72 degrees of freedom
## Multiple R-squared: 0.8333, Adjusted R-squared: 0.7916
## F-statistic:
                   20 on 18 and 72 DF, p-value: < 2.2e-16
plot(model3)
```





Fitted values Theoretical Quantiles ~ log(density) + log(taxpc) + log(prbarr) · ~ log(density) + log(taxpc) + log(prbarr) ·



Fitted values Leverage $\sim \log(\text{density}) + \log(\text{taxpc}) + \log(\text{prbarr}) \cdot \sim \log(\text{density}) + \log(\text{taxpc}) + \log(\text{prbarr}) \cdot$