# Lab 4:Crime Reduction Thoughts

Kiersten Henderson

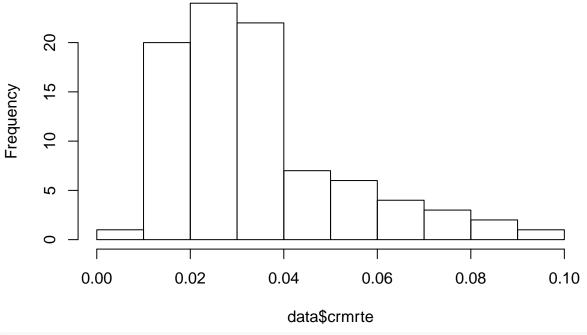
August 9, 2017

```
library(car)
library(corrplot)
library(lmtest)
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
library(sandwich)
library(stargazer)
##
## Please cite as:
   Hlavac, Marek (2015). stargazer: Well-Formatted Regression and Summary Statistics Tables.
    R package version 5.2. http://CRAN.R-project.org/package=stargazer
```

#### Transformed Dataset

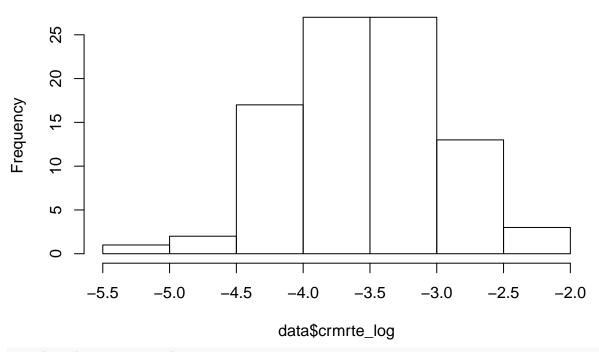
```
#copied Aaron's (checked that this incorporates what Kim did too, I added mean of wages from all categories)
data = read.csv("crime.csv")
data$crmrte_log = log(data$crmrte)
data$crmrte_1K_log = log(data$crmrte *10^3)
data$polpc_1M_log = log(data$polpc * 10^6)
data$density_1K_log = log(data$density * 10^3)
data$taxpc_log = log(data$taxpc)
data$pctmin80 = data$pctmin80 / 100
data$wtotal = (data$wcon + data$wtuc + data$wtrd + data$wfir +
              data$wser + data$wmfg + data$wfed + data$wsta + data$wloc)
data$w mean = (data$wcon + data$wtuc + data$wtrd + data$wfir +
              data$wser + data$wmfg + data$wfed + data$wsta + data$wloc)/9
sort(colnames(data))
##
    [1] "avgsen"
                          "central"
                                            "county"
                                                             "crmrte"
##
    [5] "crmrte_1K_log"
                         "crmrte_log"
                                           "density"
                                                             "density_1K_log"
##
   [9] "mix"
                          "pctmin80"
                                           "pctymle"
                                                             "polpc"
## [13] "polpc_1M_log"
                         "prbarr"
                                           "prbconv"
                                                             "prbpris"
                                           "urban"
                                                             "w mean"
## [17] "taxpc"
                         "taxpc_log"
                         "west"
                                           "wfed"
                                                             "wfir"
## [21] "wcon"
## [25] "wloc"
                          "wmfg"
                                           "wser"
                                                             "wsta"
                                                             "Х"
## [29] "wtotal"
                          "wtrd"
                                           "wtuc"
## [33] "year"
Yes, beautiful, needed that X 1K
hist(data$crmrte)
```

# Histogram of data\$crmrte



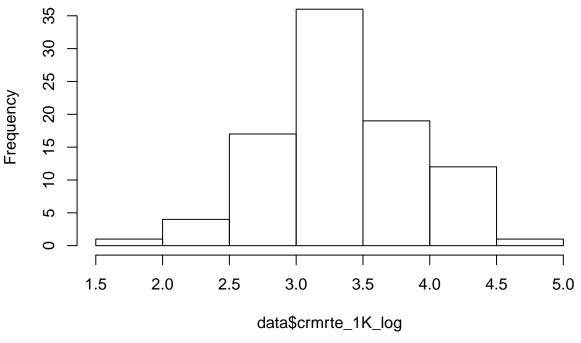
hist(data\$crmrte\_log)

# Histogram of data\$crmrte\_log



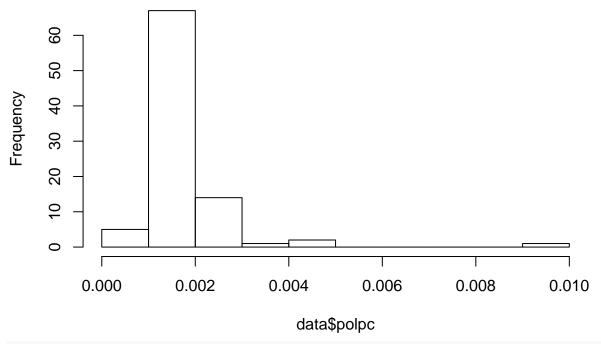
hist(data\$crmrte\_1K\_log)

# Histogram of data\$crmrte\_1K\_log



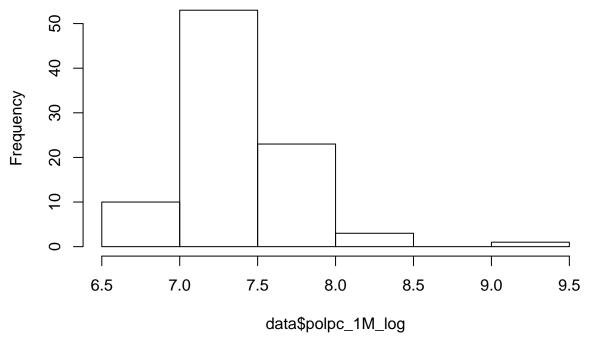
hist(data\$polpc)

# Histogram of data\$polpc



#looks better below, agreed
hist(data\$polpc\_1M\_log)

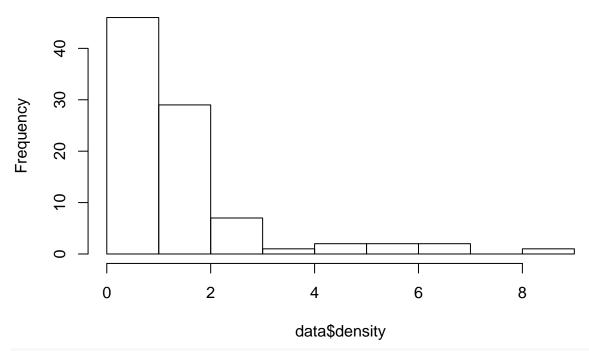
# Histogram of data\$polpc\_1M\_log



Great!

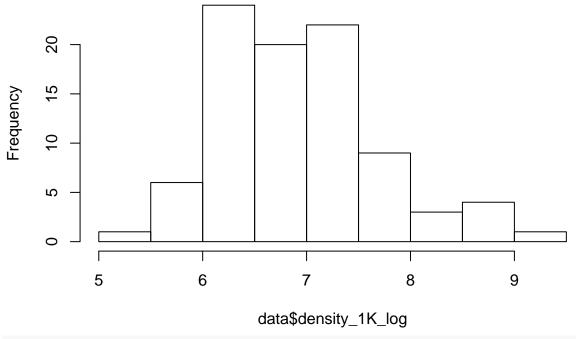
hist(data\$density)

# Histogram of data\$density



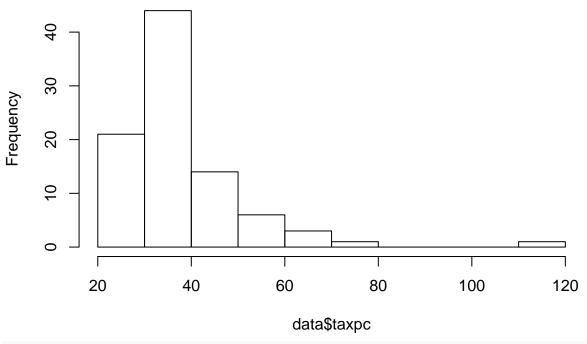
hist(data\$density\_1K\_log)

# Histogram of data\$density\_1K\_log



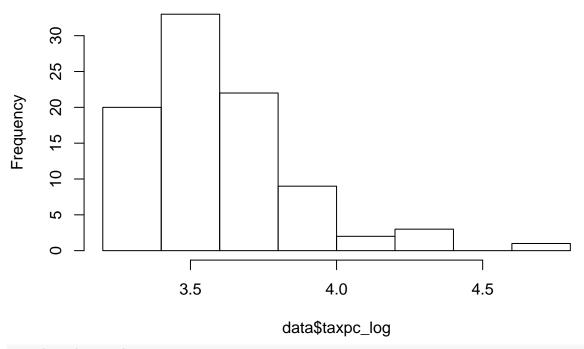
hist(data\$taxpc)

# Histogram of data\$taxpc



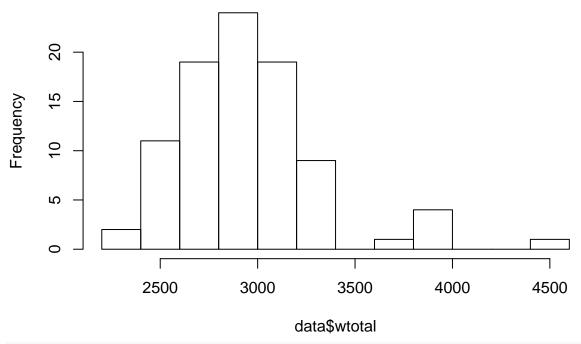
hist(data\$taxpc\_log)

# Histogram of data\$taxpc\_log



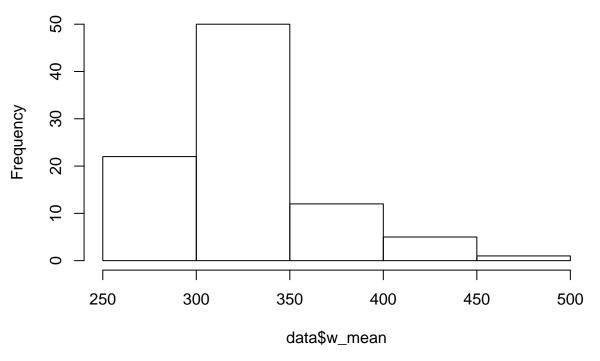
hist(data\$wtotal)

# Histogram of data\$wtotal

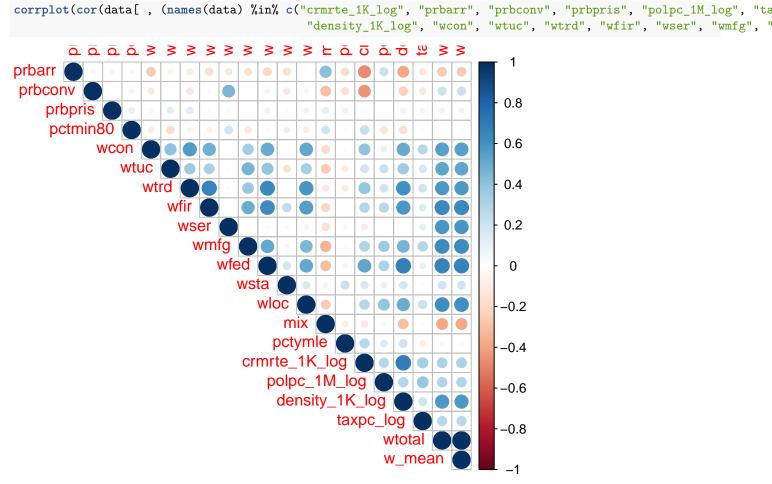


hist(data\$w\_mean)

## Histogram of data\$w\_mean



All the correlations from before the transformations hold and total wages and mean of wages from each category look identical.



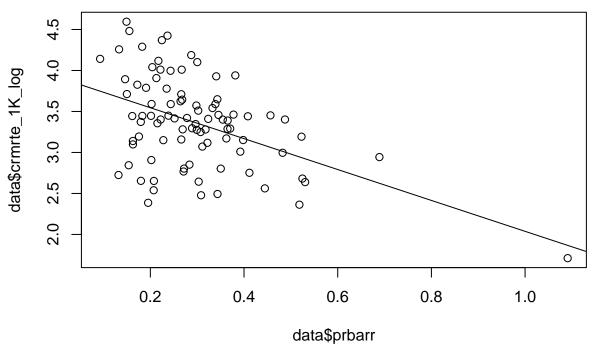
Scatterplot Matricies are too much - see individual charts below.

```
\#scatterplotMatrix(data[ , (names(data) %in% c("crmrte_1K_log", "polpc_1M_log", "prbarr", "prbconv", "prbpris $\#scatterplotMatrix(data[ , (names(data) %in% c("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ , (names(data) %in% ("crmrte_1K_log", "wcon", "wtuc", "wtrd", "wtrd", "wfir", "wser", "wmfg") $\#scatterplotMatrix(data[ (names(data) %in% (names(data)
```

It seems like the relationships between Log crime rate 1K and other variables of interest are linear.

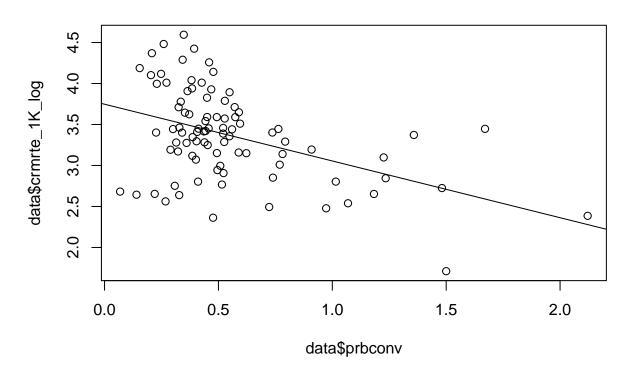
```
plot(data$prbarr,data$crmrte_1K_log, main="Log Crime rate 1K v. Prob. Arrest")
abline(lm(data$crmrte_1K_log ~ data$prbarr))
```

## Log Crime rate 1K v. Prob. Arrest

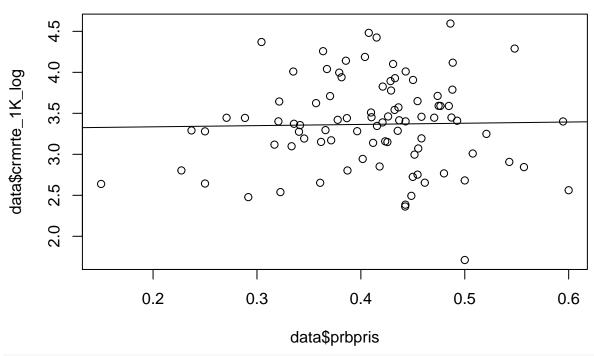


plot(data\$prbconv,data\$crmrte\_1K\_log, main="Log Crime rate 1K v.Prob. Conviction")
abline(lm(data\$crmrte\_1K\_log ~ data\$prbconv))

## Log Crime rate 1K v.Prob. Conviction

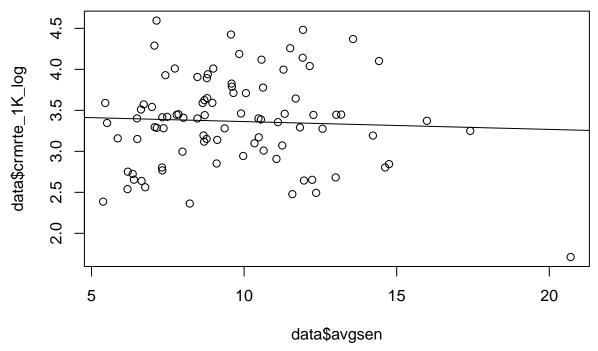


# Log Crime rate 1K v. Prob. Prison



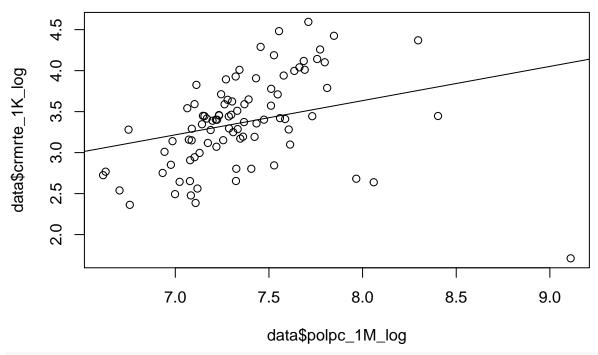
plot(data\$avgsen,data\$crmrte\_1K\_log, main="Log Crime rate 1K v. Average Sentence")
abline(lm(data\$crmrte\_1K\_log ~ data\$avgsen))

# Log Crime rate 1K v. Average Sentence



plot(data\$polpc\_1M\_log,data\$crmrte\_1K\_log, main= "Log Crime rate 1K v. Log Police per 1M Capita (R=0.28)" )
abline(lm(data\$crmrte\_1K\_log ~ data\$polpc\_1M\_log))

## Log Crime rate 1K v. Log Police per 1M Capita (R=0.28)

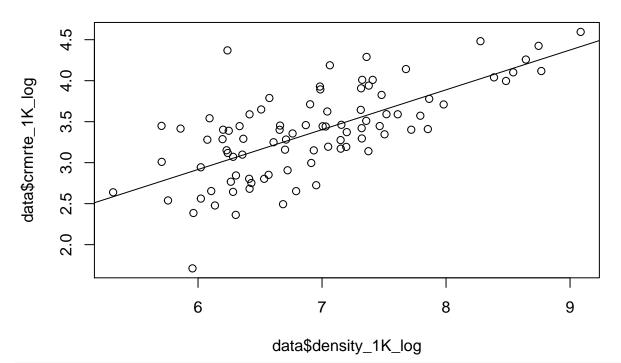


cor(data\$crmrte\_1K\_log,data\$polpc\_1M\_log)

## [1] 0.2845396

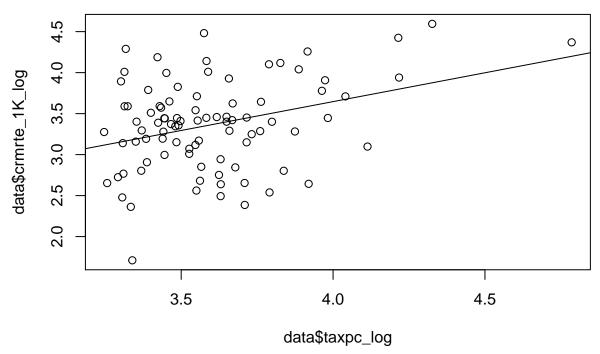
plot(data\$density\_1K\_log,data\$crmrte\_1K\_log, main="Log Crime rate 1K v. Log Density 1K")
abline(lm(data\$crmrte\_1K\_log ~ data\$density\_1K\_log))

# Log Crime rate 1K v. Log Density 1K



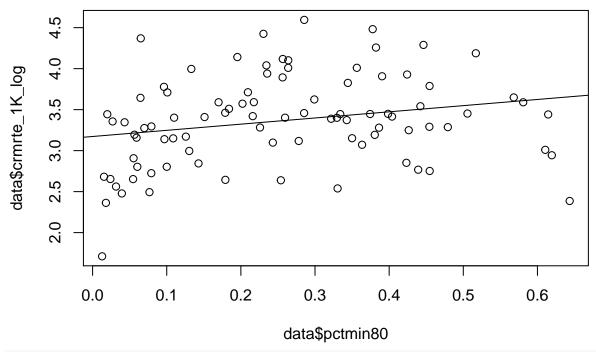
plot(data\$taxpc\_log,data\$crmrte\_1K\_log, main="Log Crime rate 1K v. Log tax per capita")
abline(lm(data\$crmrte\_1K\_log ~ data\$taxpc\_log))

## Log Crime rate 1K v. Log tax per capita



plot(data\$pctmin80,data\$crmrte\_1K\_log, main="Log Crime rate 1K v. Percent Minority, (R=0.23)")
abline(lm(data\$crmrte\_1K\_log ~ data\$pctmin80))

# Log Crime rate 1K v. Percent Minority, (R=0.23)

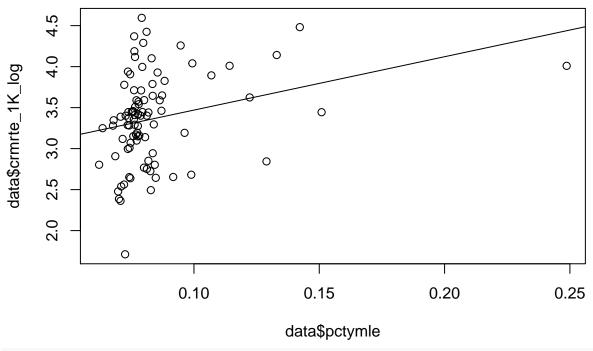


cor(data\$crmrte\_1K\_log,data\$pctmin80 )

## [1] 0.2329182

plot(data\$pctymle,data\$crmrte\_1K\_log, main="Log Crime rate 1K v. Percent Young Male, (R=0.28)")
abline(lm(data\$crmrte\_1K\_log ~ data\$pctymle))

# Log Crime rate 1K v. Percent Young Male, (R=0.28)

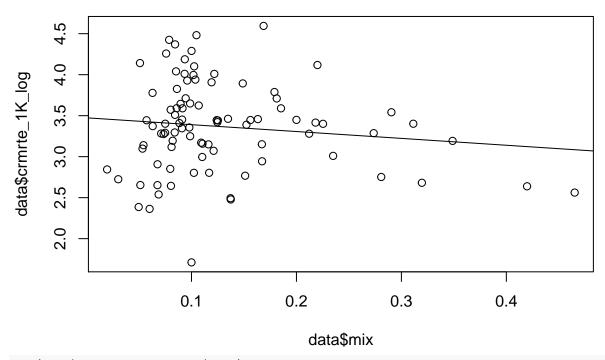


cor(data\$crmrte\_1K\_log,data\$pctymle )

## [1] 0.2781547

plot(data\$mix,data\$crmrte\_1K\_log, main="Log Crime rate 1K v. Offense Mix, (R=-0.12)")
abline(lm(data\$crmrte\_1K\_log ~ data\$mix))

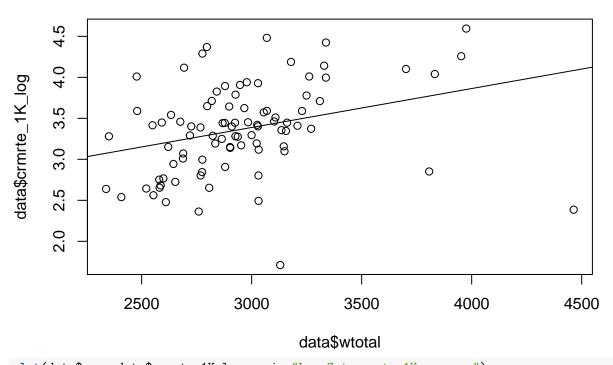
# Log Crime rate 1K v. Offense Mix, (R=-0.12)



cor(data\$crmrte\_1K\_log,data\$mix )

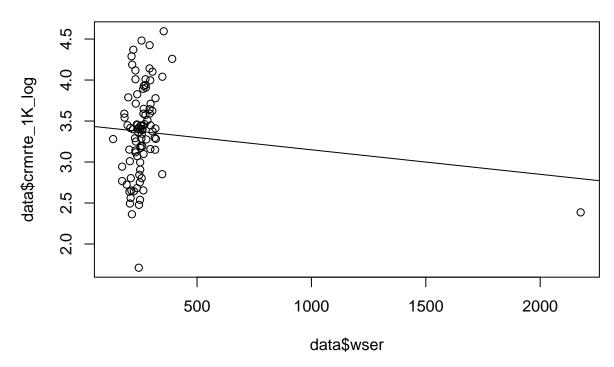
## [1] -0.1247345

## Log Crime rate 1K v. Total Wages



plot(data\$wser,data\$crmrte\_1K\_log, main="Log Crime rate 1K v. wser")
abline(lm(data\$crmrte\_1K\_log ~ data\$wser))

# Log Crime rate 1K v. wser



Modeling conclusions:

The best way to look at models is to go to the stargazer summaries because the far left column tells you what variables are included..there's one on line 200 in RMD..and one on line 256.

-we can achieve adjusted R squared of 0.84 -surprisingly, wages dont seem to matter for model anymore (i did do linear Hypothesis test) -density can surprisingly be excluded (you can get 0.75 ish with exclusion of density), but its probably better to include it -if you exclude density, then the wages categories become significant and bump model to 0.78 R2 (but the individual contributions of wages are not practically significant at all)

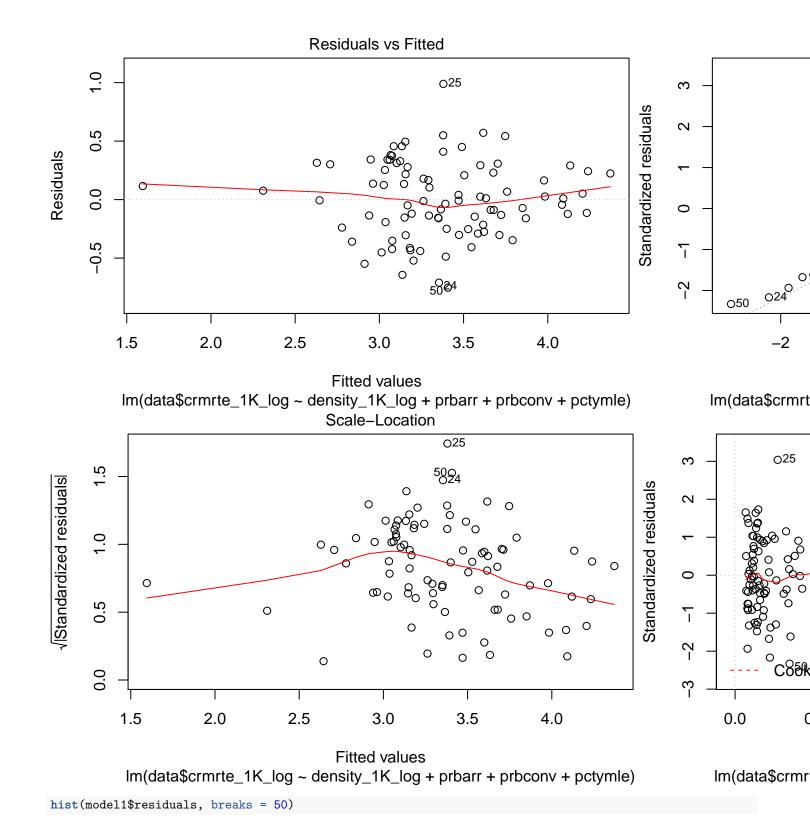
Further messing around to be certain but preliminary findings:

-percent young male doesnt matter -seems like models benefit quite a bit from including police per capita and both prob arrest and conviction

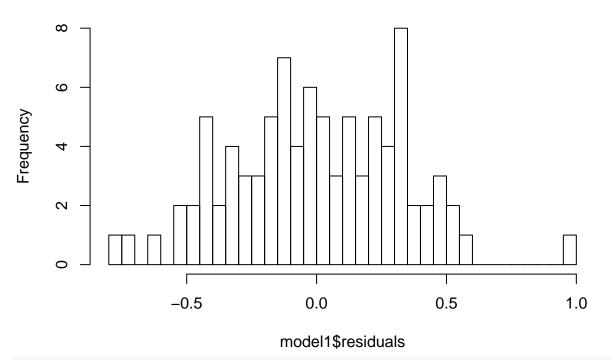
Right now i'm leaning towards model6

```
model1 = lm(data$crmrte_1K_log ~ density_1K_log+prbarr +prbconv+ pctymle, data = data)
summary(model1)
##
## Call:
## lm(formula = data$crmrte_1K_log ~ density_1K_log + prbarr + prbconv +
       pctymle, data = data)
##
##
## Residuals:
##
       Min
                  1Q
                       Median
                                     3Q
                                             Max
## -0.75303 -0.23287 -0.00751 0.25071 0.98956
##
## Coefficients:
##
                  Estimate Std. Error t value Pr(>|t|)
```

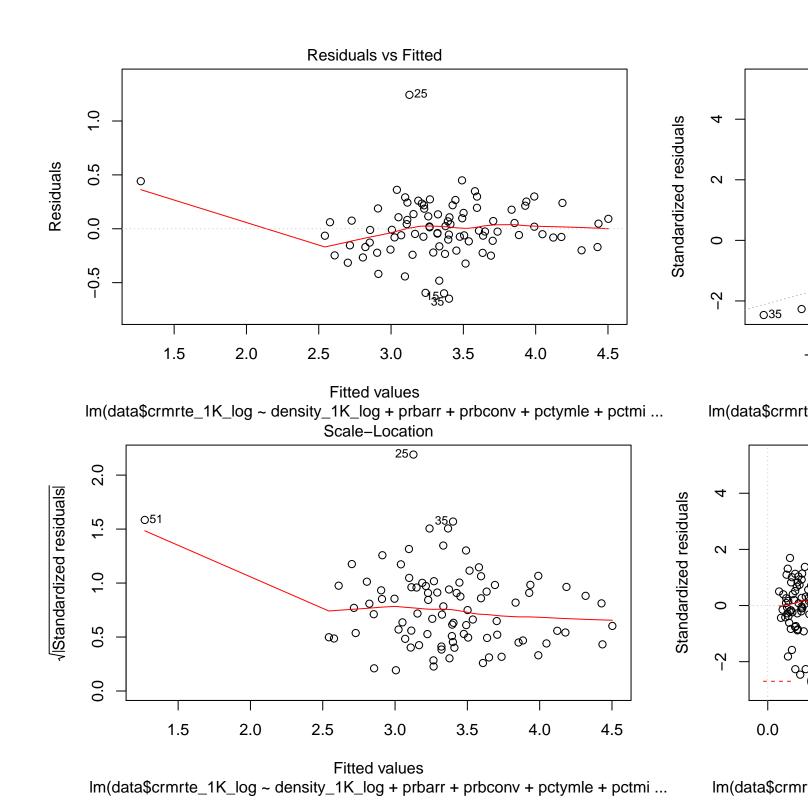
```
0.4329
                                      3.455 0.000861 ***
## (Intercept)
                 1.4958
## density_1K_log
                 0.3411
                              0.0515
                                       6.624 3.01e-09 ***
                              0.2857
                                     -4.069 0.000105 ***
## prbarr
                  -1.1624
## prbconv
                  -0.5230
                              0.1055
                                     -4.957 3.61e-06 ***
## pctymle
                  1.6596
                              1.5716
                                      1.056 0.293987
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3347 on 85 degrees of freedom
## Multiple R-squared: 0.6446, Adjusted R-squared: 0.6279
## F-statistic: 38.55 on 4 and 85 DF, p-value: < 2.2e-16
```



## Histogram of model1\$residuals

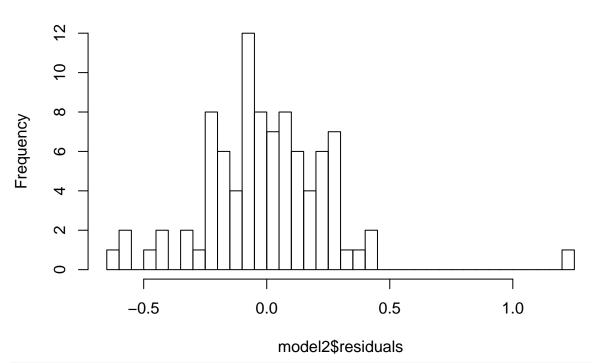


```
coeftest(model1, vcov = vcovHC)
##
## t test of coefficients:
##
##
                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                  1.495837
                             0.501070 2.9853 0.0036995 **
                             0.053859 6.3336 1.091e-08 ***
## density_1K_log 0.341124
                 -1.162418
## prbarr
                             0.295280 -3.9367 0.0001685 ***
## prbconv
                 -0.523006
                             0.118806 -4.4022 3.098e-05 ***
## pctymle
                  1.659561
                             0.946738 1.7529 0.0832203 .
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
#vcovHC(model1)
(se.model1 = sqrt(diag(vcovHC(model1))))
##
      (Intercept) density_1K_log
                                         prbarr
                                                       prbconv
                                                                      pctymle
##
      0.50106995
                      0.05385929
                                    0.29528008
                                                    0.11880569
                                                                   0.94673807
Below, added in minority
model2 = lm(data$crmrte_1K_log ~ density_1K_log+prbarr +prbconv+ pctymle+pctmin80, data = data)
#summary(model2)
plot(model2)
```



hist(model2\$residuals, breaks = 50)

## Histogram of model2\$residuals

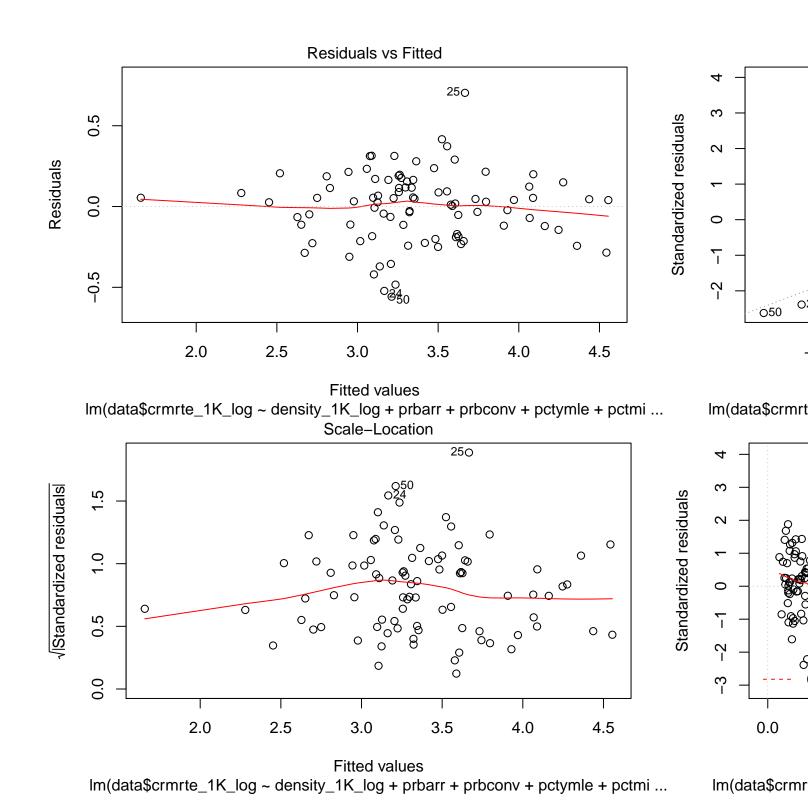


```
coeftest(model2, vcov = vcovHC)
##
```

```
## t test of coefficients:
##
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                  0.89481
                             0.66494 1.3457 0.182024
## density_1K_log 0.38621
                             0.05763 6.7016 2.218e-09 ***
## prbarr
                 -1.14180
                             0.62444 -1.8285 0.071019 .
## prbconv
                 -0.53618
                             0.16358 -3.2778  0.001523 **
## pctymle
                  1.50512
                             0.73932 2.0358 0.044920 *
## pctmin80
                  1.17823
                             0.22387 5.2631 1.068e-06 ***
##
                  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
#vcovHC(model2)
se.model2 = sqrt(diag(vcovHC(model2)))
```

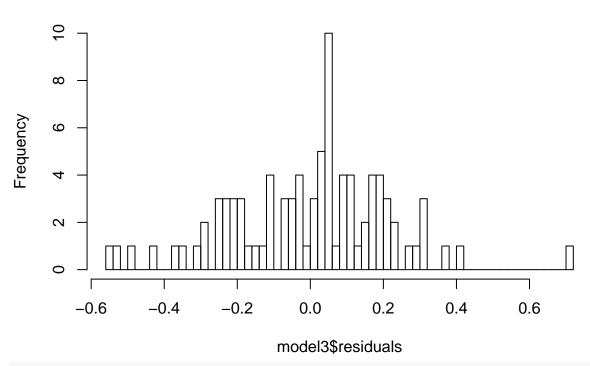
This model below is model2 with the transformed polic per capita

```
model3 = lm(data$crmrte_1K_log ~ density_1K_log+prbarr +prbconv+ pctymle+pctmin80+polpc_1M_log, data = data)
#summary(model3)
plot(model3)
```



hist(model3\$residuals, breaks = 50)

## Histogram of model3\$residuals

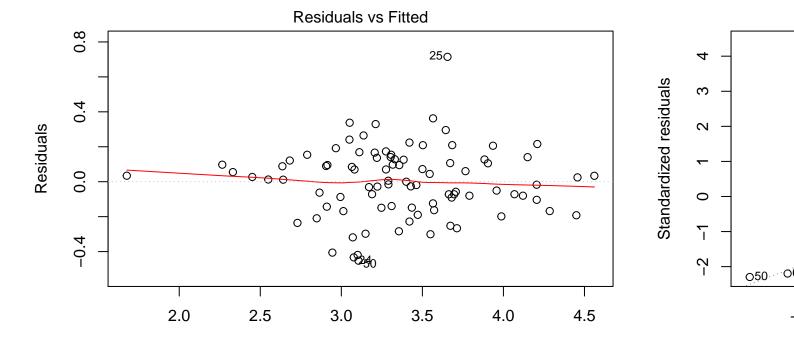


```
coeftest(model3, vcov = vcovHC)
```

```
##
## t test of coefficients:
##
##
                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 -1.497286
                             0.600865 -2.4919 0.014697 *
## density_1K_log 0.291274
                             0.064846 4.4918 2.263e-05 ***
## prbarr
                 -1.663389
                             0.312333 -5.3257 8.438e-07 ***
## prbconv
                 -0.607604
                             0.107197 -5.6681 2.036e-07 ***
## pctymle
                  0.418323
                             2.096464 0.1995 0.842331
## pctmin80
                             0.152906 8.3343 1.402e-12 ***
                  1.274366
## polpc_1M_log
                  0.449832
                             0.139652 3.2211 0.001825 **
##
## Signif. codes:
                  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#vcovHC(model3)
se.model3 = sqrt(diag(vcovHC(model3)))
```

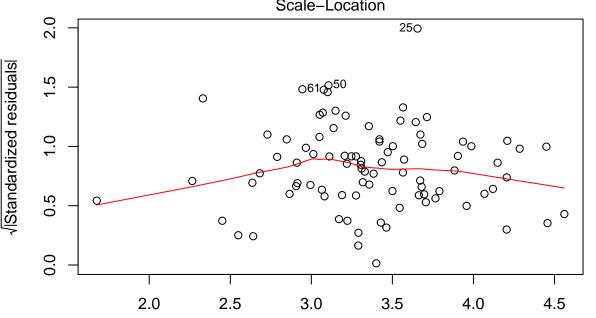
This model is model 3 with all the kinds of employment included:

```
model4 = lm(data$crmrte_1K_log ~ density_1K_log+prbarr +prbconv+ pctymle+pctmin80+polpc_1M_log+wcon+wtuc+wtrd
#summary(model4)
plot(model4)
```



Fitted values
Im(data\$crmrte\_1K\_log ~ density\_1K\_log + prbarr + prbconv + pctymle + pctmi ...
Scale-Location

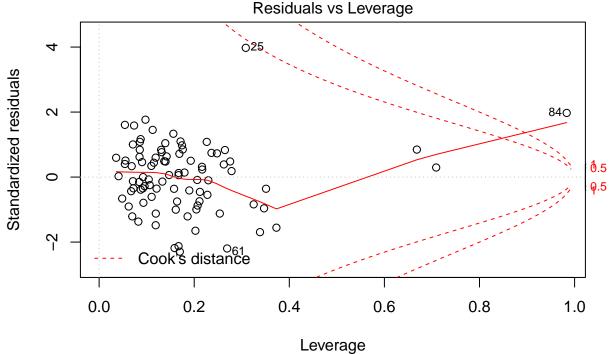
Im(data\$crmrt



Fitted values Im(data\$crmrte\_1K\_log ~ density\_1K\_log + prbarr + prbconv + pctymle + pctmi ...

## Warning in sqrt(crit \* p \* (1 - hh)/hh): NaNs produced

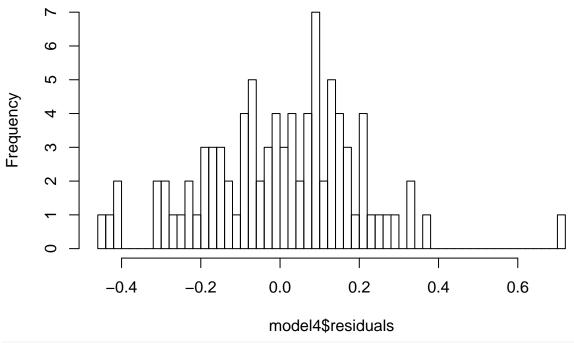
## Warning in sqrt(crit \* p \* (1 - hh)/hh): NaNs produced



Im(data\$crmrte\_1K\_log ~ density\_1K\_log + prbarr + prbconv + pctymle + pctmi ...

hist(model4\$residuals, breaks = 50)

## Histogram of model4\$residuals



```
coeftest(model4, vcov = vcovHC)
```

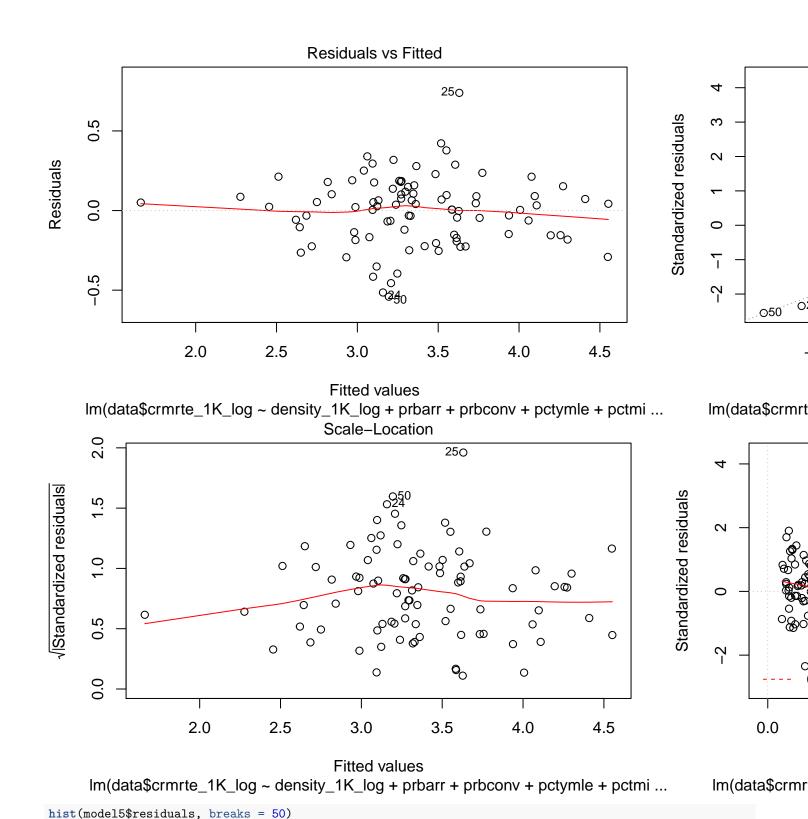
```
## prbconv
                -5.2976e-01 1.2117e-01 -4.3720 3.960e-05 ***
## pctymle
                  1.1725e+00 1.6036e+00 0.7312 0.4669715
                1.3395e+00 1.6181e-01 8.2783 3.854e-12 ***
## pctmin80
## polpc_1M_log 4.6263e-01 1.7587e-01 2.6305 0.0103679 *
          5.6843e-04 (.5051e 04 0.1078 0.9144220
4.4261e-05 4.1046e-04 0.1078 0.9144220
## wcon
              4.4261e-05 4.1046e-04 0.1078 0.9144220 6.5539e-04 1.0947e-03 0.5987 0.5511983 -1.5050e-03 8.4022e-04 -1.7912 0.0773540 .
## wtuc
## wtrd
## wfir
## wmfg
                -7.1987e-05 3.4695e-04 -0.2075 0.8362005
                  6.6007e-04 9.5917e-04 0.6882 0.4934964
## wfed
## wsta
                  -9.7119e-04 6.7006e-04 -1.4494 0.1514501
## wloc
                  7.0616e-04 1.7180e-03 0.4110 0.6822275
## wser
               -2.4325e-04 1.6258e-03 -0.1496 0.8814731
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
#vcovHC(model4)
se.model4 = sqrt(diag(vcovHC(model4)))
linearHypothesis(model4, c("wcon = 0", "wtuc = 0", "wtrd = 0", "wfir = 0", "wmfg = 0", "wfed = 0", "wsta = 0"
## Linear hypothesis test
##
## Hypothesis:
## wcon = 0
## wtuc = 0
## wtrd = 0
## wfir = 0
## wmfg = 0
## wfed = 0
## wsta = 0
## wloc = 0
## Model 1: restricted model
## Model 2: data$crmrte_1K_log ~ density_1K_log + prbarr + prbconv + pctymle +
       pctmin80 + polpc_1M_log + wcon + wtuc + wtrd + wfir + wmfg +
##
       wfed + wsta + wloc + wser
##
## Note: Coefficient covariance matrix supplied.
##
##
   Res.Df Df
                   F Pr(>F)
## 1
         82
## 2
         74 8 0.9856 0.4542
stargazer(model1, model2, model3, model4,type="text",
se=list(se.model1, se.model2, se.model3, se.model4), star.cutoffs=c(0.05, 0.01, 0.001))
##
```

```
##
                                           Dependent variable:
##
##
                                             crmrte_1K_log
                       (1)
                                        (2)
                                                         (3)
                                                                           (4)
## density_1K_log
                      0.341***
                                       0.386***
                                                        0.291***
                                                                         0.280***
##
                     (0.054)
                                       (0.058)
                                                        (0.065)
                                                                          (0.070)
##
                     -1.162***
                                                       -1.663***
                                                                         -1.740***
## prbarr
                                       -1.142
##
                     (0.295)
                                      (0.624)
                                                       (0.312)
                                                                         (0.343)
##
                     -0.523***
## prbconv
                                       -0.536**
                                                       -0.608***
                                                                         -0.530***
```

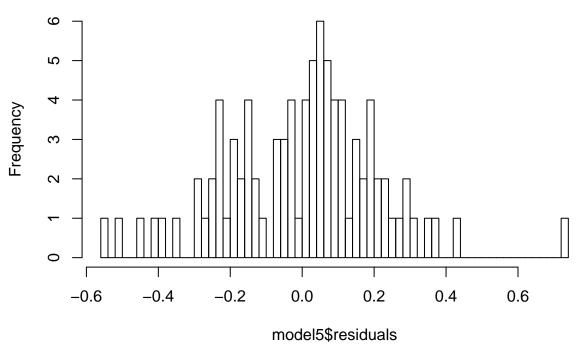
		<b>4</b>	4		4
##		(0.119)	(0.164)	(0.107)	(0.121)
##		1 660	1 FOE:	0.418	1 170
##	pctymle	1.660 (0.947)	1.505* (0.739)	(2.096)	1.173 (1.604)
##		(0.947)	(0.739)	(2.096)	(1.004)
	pctmin80		1.178***	1.274***	1.340***
##	рссштноо		(0.224)	(0.153)	(0.162)
##			(0.224)	(0.155)	(0.102)
	polpc_1M_log			0.450**	0.463**
##	Po-10-111-108			(0.140)	(0.176)
##				(0.110)	(0.2.0)
	wcon				0.001
##					(0.001)
##					,
##	wtuc				0.00004
##					(0.0004)
##					
##	wtrd				0.001
##					(0.001)
##					
##	wfir				-0.002
##					(0.001)
##					
	wmfg				-0.0001
##					(0.0003)
##					
	wfed				0.001
##					(0.001)
##					0.004
	wsta				-0.001
##					(0.001)
##					0.001
##	wloc				(0.002)
##					(0.002)
	wser				-0.0002
##	MPGT				(0.002)
##					(0.002)
	Constant	1.496**	0.895	-1.497*	-1.524
##		(0.501)	(0.665)	(0.601)	(0.823)
##			- · · · · ·		
##					
##	Observations	90	90	90	90
##	R2	0.645	0.774	0.844	0.871
##	Adjusted R2	0.628	0.760	0.832	0.845
##	Residual Std. Erre	or 0.335 (df = 85)			
##	F Statistic	38.549*** (df = 4; 85)	57.386*** (df = 5; 84)	74.695*** (df = 6; 83)	33.307*** (df = 1)
##	Note:			*p<0	.05; **p<0.01; ***p

This model below (5) is model with just federal worker wage included.

```
model5 = lm(data$crmrte_1K_log ~ density_1K_log+prbarr +prbconv+ pctymle+pctmin80+polpc_1M_log+wfed, data = d
#summary(model5)
plot(model5)
```



## Histogram of model5\$residuals

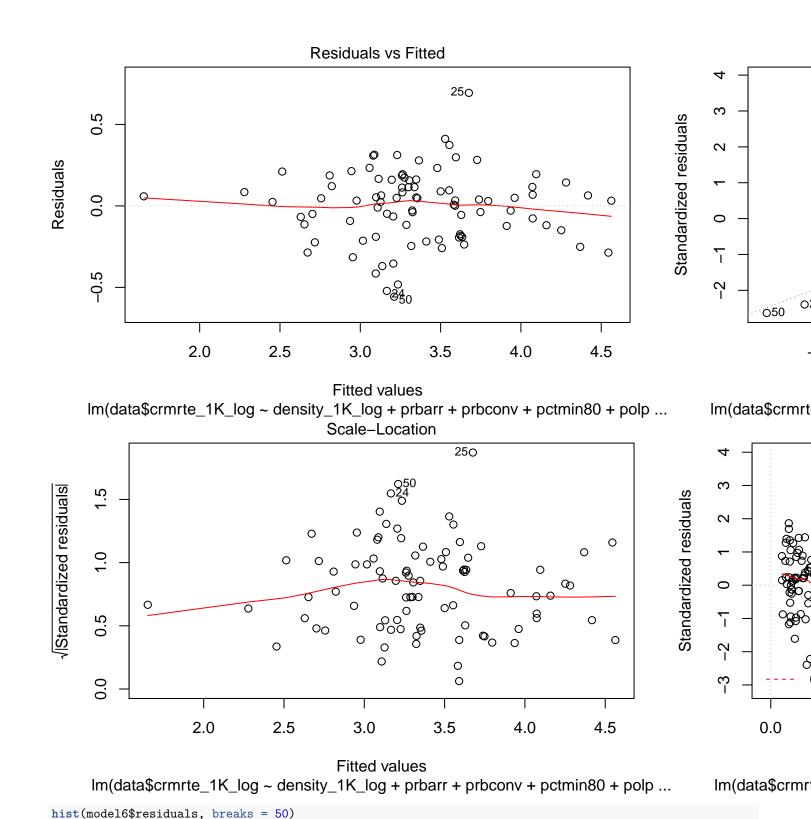


```
coeftest(model5, vcov = vcovHC)
```

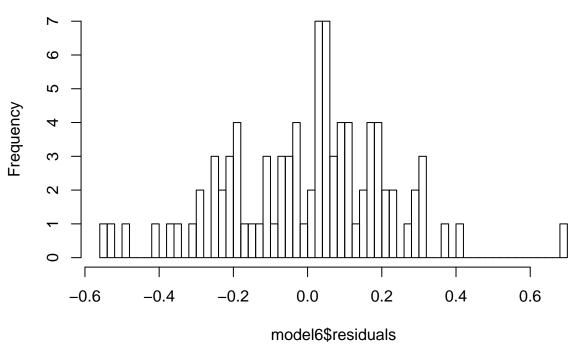
```
##
## t test of coefficients:
##
##
                 Estimate Std. Error t value Pr(>|t|)
               -1.44931653 0.65880208 -2.1999 0.030626 *
## (Intercept)
## density_1K_log 0.26639476 0.05545651 4.8037 6.934e-06 ***
## prbarr
               ## prbconv
               ## pctymle
               0.67089631 2.40371156
                                    0.2791 0.780864
                                    8.4241 1.007e-12 ***
## pctmin80
               1.24720499 0.14805278
## polpc_1M_log
               0.43809216 0.16483422 2.6578
                                           0.009454 **
## wfed
                0.00044852 0.00086332 0.5195
                                           0.604790
##
               0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
#vcovHC(model5)
se.model5 = sqrt(diag(vcovHC(model5)))
```

This model below (6) is model3 with percent young male excluded.

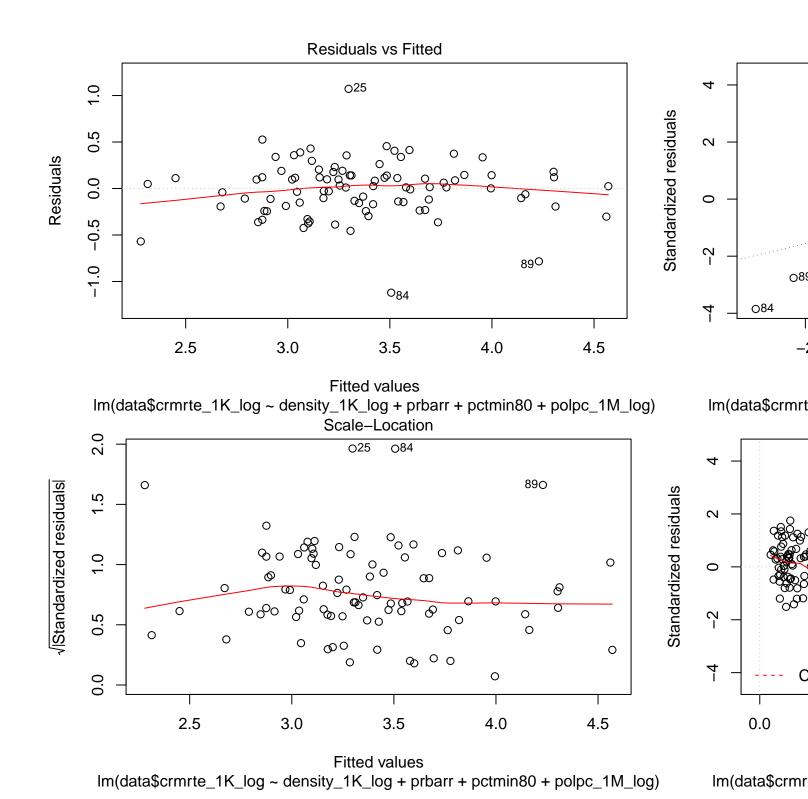
```
model6 = lm(data$crmrte_1K_log ~ density_1K_log+prbarr +prbconv+pctmin80+polpc_1M_log, data = data)
#summary(model6)
plot(model6)
```



## Histogram of model6\$residuals

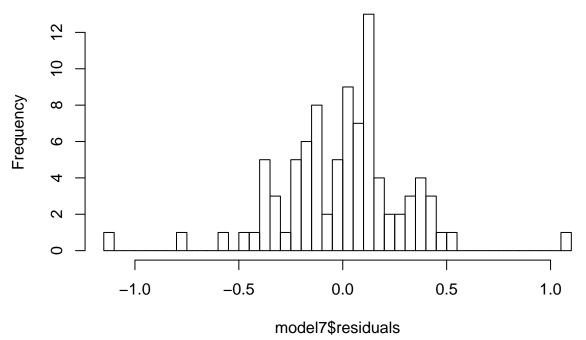


```
coeftest(model6, vcov = vcovHC)
##
## t test of coefficients:
##
##
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                  -1.493838
                              0.600417 -2.4880 0.0148224 *
## density_1K_log 0.291715
                              0.063965 4.5605 1.722e-05 ***
## prbarr
                  -1.678919
                              0.285219 -5.8864 7.848e-08 ***
## prbconv
                              0.102938 -5.9475 6.040e-08 ***
                  -0.612226
## pctmin80
                   1.276361
                              0.151828 8.4066 9.274e-13 ***
                              0.129769 3.5033 0.0007391 ***
  polpc_1M_log
                   0.454625
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
#vcovHC(model6)
se.model6 = sqrt(diag(vcovHC(model6)))
model7 = lm(data$crmrte_1K_log ~ density_1K_log+prbarr +pctmin80+polpc_1M_log, data = data)
#summary(model7)
plot(model7)
```



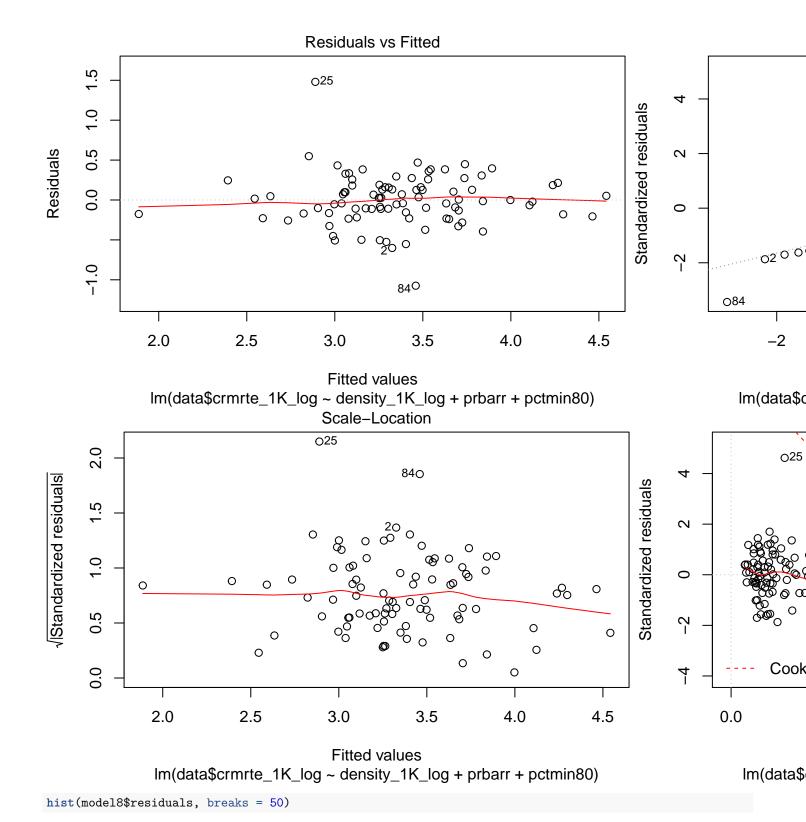
hist(model7\$residuals, breaks = 50)

# Histogram of model7\$residuals

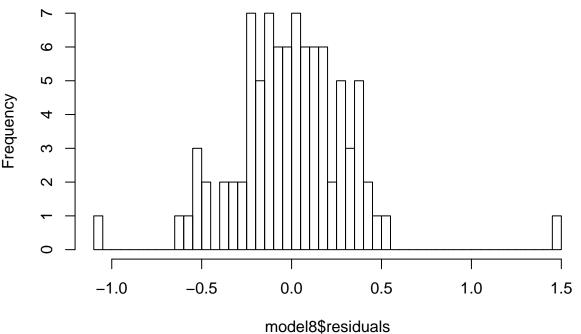


plot(model8)

#### coeftest(model7, vcov = vcovHC) ## ## t test of coefficients: ## ## Estimate Std. Error t value Pr(>|t|)## (Intercept) -1.975460 1.417207 -1.3939 0.074566 5.2544 1.085e-06 \*\*\* ## density\_1K\_log 0.391800 0.618979 -2.1295 ## prbarr -1.318135 0.0361 \* ## pctmin80 1.233537 0.269461 4.5778 1.592e-05 \*\*\* ## polpc\_1M\_log 0.367069 0.233681 1.5708 0.1199 ## ## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1 #vcovHC(model7) se.model7 = sqrt(diag(vcovHC(model7))) model8 = lm(data\$crmrte\_1K\_log ~ density\_1K\_log+prbarr +pctmin80, data = data) #summary(model8)



## Histogram of model8\$residuals



```
coeftest(model8, vcov = vcovHC)
## t test of coefficients:
##
##
               Estimate Std. Error t value Pr(>|t|)
              ## (Intercept)
                       0.05438 8.5643 3.793e-13 ***
## density_1K_log 0.46573
## prbarr
             ## pctmin80
               ## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
#vcovHC(model8)
se.model8 = sqrt(diag(vcovHC(model8)))
stargazer(model1, model2, model3, model6, model7, model8,type="text",
se=list(se.model1, se.model2, se.model3, se.model6, se.model7, se.model8), star.cutoffs=c(0.05, 0.01, 0.001))
##
                                                                      Dependent variable:
##
##
                                                                         crmrte_1K_log
                           (1)
                                                                   (3)
##
## density_1K_log
                         0.341***
                                             0.386***
                                                                 0.291***
                                                                                     0.292***
##
                         (0.054)
                                             (0.058)
                                                                 (0.065)
                                                                                     (0.064)
##
                        -1.162***
                                                                -1.663***
                                                                                    -1.679***
## prbarr
                                             -1.142
##
                         (0.295)
                                             (0.624)
                                                                 (0.312)
                                                                                     (0.285)
##
## prbconv
                        -0.523***
                                             -0.536**
                                                                -0.608***
                                                                                    -0.612***
```

(0.164)

1.505\*

(0.107)

0.418

(0.103)

(0.119)

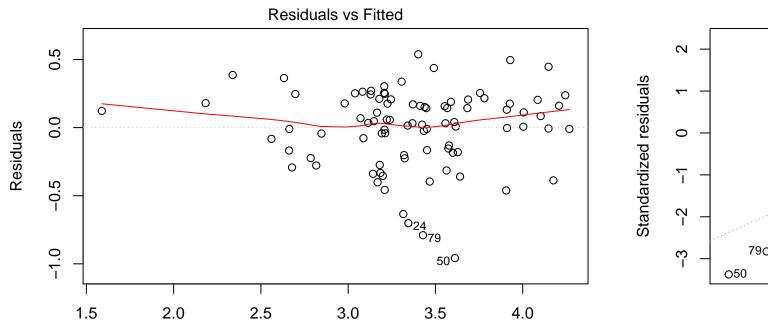
1.660

##

##

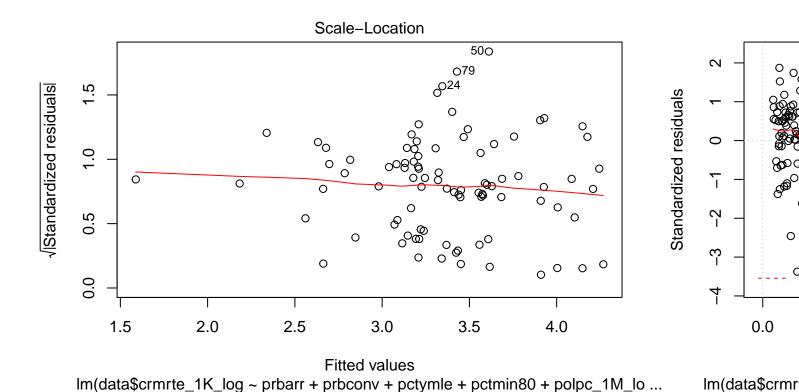
## pctymle

```
(0.947)
                                                         (0.739)
##
                                                                                  (2.096)
##
                                                         1.178***
                                                                                  1.274***
                                                                                                          1.276***
## pctmin80
##
                                                         (0.224)
                                                                                  (0.153)
                                                                                                          (0.152)
##
                                                                                  0.450**
                                                                                                          0.455***
   polpc_1M_log
                                                                                  (0.140)
                                                                                                          (0.130)
##
##
                                1.496**
                                                          0.895
                                                                                  -1.497*
##
   Constant
                                                                                                          -1.494*
##
                                (0.501)
                                                         (0.665)
                                                                                  (0.601)
                                                                                                          (0.600)
##
##
                                   90
                                                                                     90
                                                                                                              90
##
   Observations
                                                            90
                                                          0.774
## R2
                                 0.645
                                                                                   0.844
                                                                                                           0.843
## Adjusted R2
                                 0.628
                                                          0.760
                                                                                   0.832
                                                                                                           0.834
## Residual Std. Error
                           0.335 (df = 85)
                                                    0.269 (df = 84)
                                                                             0.225 (df = 83)
                                                                                                      0.223 \text{ (df = 84)}
## F Statistic
                        38.549*** (df = 4; 85) 57.386*** (df = 5; 84) 74.695*** (df = 6; 83) 90.516*** (df = 5
## Note:
model9 = lm(data$crmrte_1K_log ~ prbarr +prbconv+ pctymle+pctmin80+polpc_1M_log, data = data)
#summary(model9)
plot(model9)
```



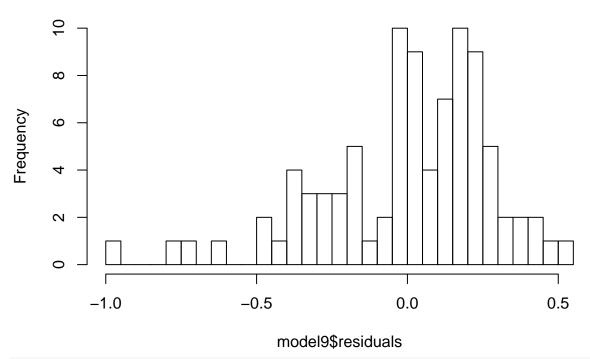
Fitted values
Im(data\$crmrte\_1K\_log ~ prbarr + prbconv + pctymle + pctmin80 + polpc\_1M\_lo ...

Im(data\$crmrt



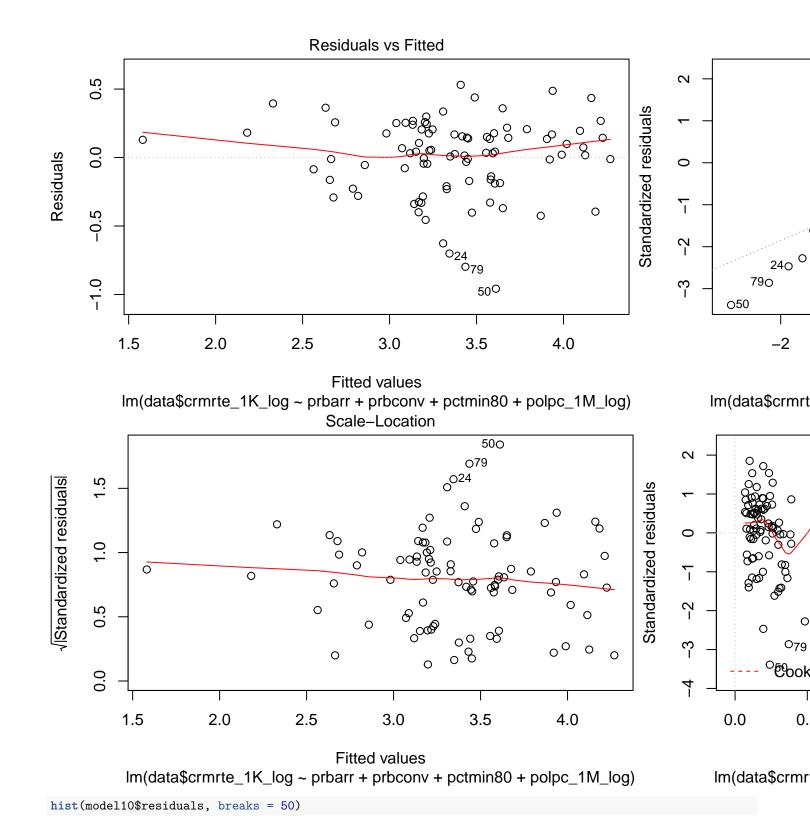
hist(model9\$residuals, breaks = 50)

## Histogram of model9\$residuals

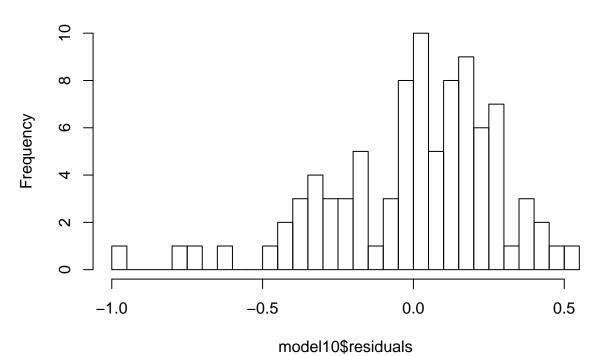


```
coeftest(model9, vcov = vcovHC)
```

```
## pctymle
               0.65841 2.53554 0.2597
                                           0.7958
## pctmin80
             ## polpc_1M_log 0.67744 0.10306 6.5731 3.928e-09 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
#vcovHC(model9)
se.model9 = sqrt(diag(vcovHC(model9)))
stargazer(model3, model6, model7, model8, model9, type="text",
se=list(se.model3, se.model6, se.model7, se.model8, se.model9), star.cutoffs=c(0.05, 0.01, 0.001))
##
##
                                                                Dependent variable:
##
##
                                                                  crmrte_1K_log
##
                            (1)
                                                (2)
## -----
                           0.291***
                                                0.292***
                                                                     0.392***
                                                                                          0.466***
## density_1K_log
##
                           (0.065)
                                                (0.064)
                                                                     (0.075)
                                                                                          (0.054)
##
                                                                     -1.318*
## prbarr
                          -1.663***
                                               -1.679***
                                                                                          -0.936**
##
                          (0.312)
                                               (0.285)
                                                                     (0.619)
                                                                                          (0.303)
##
                                               -0.612***
## prbconv
                          -0.608***
##
                           (0.107)
                                                (0.103)
##
                           0.418
## pctymle
##
                           (2.096)
##
                           1.274***
                                                1.276***
                                                                     1.234***
                                                                                          1.159***
## pctmin80
##
                           (0.153)
                                                (0.152)
                                                                     (0.269)
                                                                                          (0.270)
##
## polpc_1M_log
                           0.450**
                                                0.455***
                                                                      0.367
##
                           (0.140)
                                                (0.130)
                                                                     (0.234)
##
## Constant
                           -1.497*
                                                -1.494*
                                                                      -1.975
                                                                                           0.120
##
                           (0.601)
                                                (0.600)
                                                                                          (0.477)
                                                                     (1.417)
                            90
## Observations
                                                  90
                                                                        90
                                                                                             90
## R2
                           0.844
                                                 0.843
                                                                     0.703
                                                                                           0.654
## Adjusted R2
                           0.832
                                                 0.834
                                                                      0.689
                                                                                           0.642
## Residual Std. Error 0.225 (df = 83) 0.223 (df = 84) 0.306 (df = 85)
                                                                                      0.329 (df = 86
## F Statistic 74.695*** (df = 6; 83) 90.516*** (df = 5; 84) 50.216*** (df = 4; 85) 54.096*** (df = 3
## Note:
model10 = lm(data$crmrte_1K_log ~ prbarr +prbconv+pctmin80+polpc_1M_log, data = data)
#summary(model10)
plot(model10)
```



## Histogram of model10\$residuals



(1)

0.291\*\*\*

(0.065)

-1.663\*\*\*

(0.312)

-0.608\*\*\*

(0.107)

```
coeftest(model10, vcov = vcovHC)
```

##

## ##

##

##

##

##

##

##

##

##

## prbarr

## prbconv

density\_1K\_log

```
## t test of coefficients:
##
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.822437 0.736426 -1.1168
             ## prbarr
## prbconv
             ## pctmin80
             1.171711
                      0.185401 6.3199 1.158e-08 ***
## polpc_1M_log 0.685529
                      0.097882 7.0036 5.462e-10 ***
##
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#vcovHC(model10)
se.model10 = sqrt(diag(vcovHC(model10)))
stargazer(model3, model6, model7, model8, model9, model10,type="text",
se=list(se.model3, se.model6, se.model7, se.model8, se.model9, se.model10), star.cutoffs=c(0.05, 0.01, 0.001))
##
```

Dependent variable:

crmrte\_1K\_log

(4)

0.466\*\*\*

(0.054)

-0.936\*\*

(0.303)

(3)

0.392\*\*\*

(0.075)

-1.318\*

(0.619)

37

(2)

0.292\*\*\*

(0.064)

-1.679\*\*\*

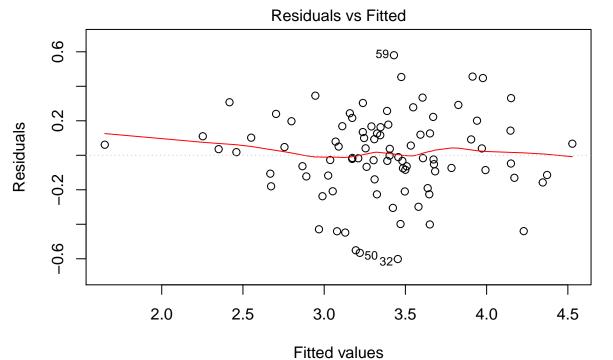
(0.285)

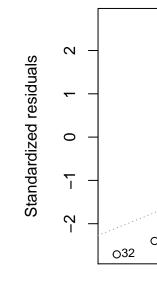
-0.612\*\*\*

(0.103)

	0.418			
	(2.096)			
				ļ
	1.274***	1.276***	1.234***	1.159***
	(0.153)	(0.152)	(0.269)	(0.270)
_log	0.450**	0.455***	0.367	ļ
_	(0.140)	(0.130)	(0.234)	
	-1.497*	-1.494*	-1.975	0.120
	(0.601)	(0.600)	(1.417)	(0.477)
ions	90	90	90	90
	0.844	0.843	0.703	0.654
R2	0.832	0.834	0.689	0.642
Std. Error	0.225 (df = 83)	0.223 (df = 84)	0.306 (df = 85)	0.329 (df = 86)
tic	74.695*** (df = 6: 8'	3) 90.516*** (df = 5; 84)	50.216*** (df = 4; 85)	54.096*** (df = 3
- i	ions R2 Std. Error	(2.096)  1.274*** (0.153)  log	(2.096)  1.274***	(2.096)  1.274***

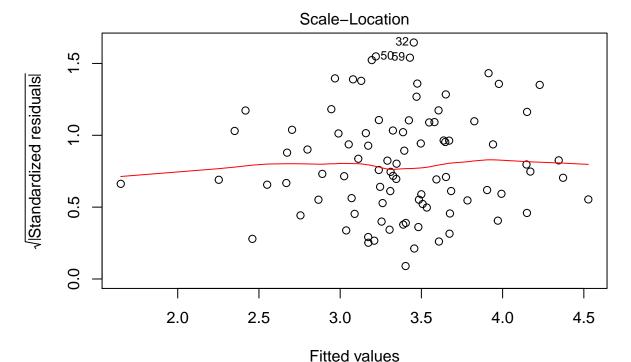
model11 = lm(data\$crmrte\_1K\_log ~ prbarr +prbconv+pctmin80+polpc\_1M\_log+wcon+wtuc+wtrd+wfir+wmfg+wfed+wsta+wl
#summary(model11)
plot(model11)





Im(data\$crmrte\_1K\_log ~ prbarr + prbconv + pctmin80 + polpc\_1M\_log + wcon + ...

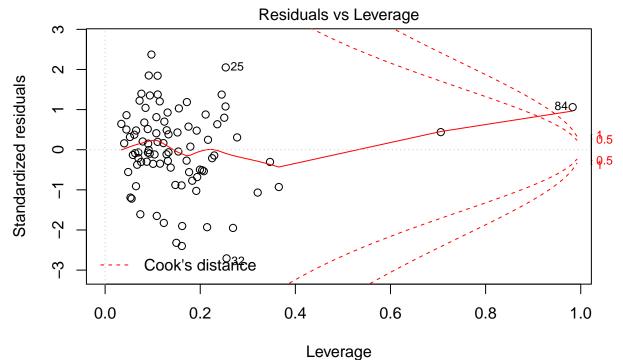
lm(data\$crmrt



Im(data\$crmrte\_1K\_log ~ prbarr + prbconv + pctmin80 + polpc\_1M\_log + wcon + ...

## Warning in sqrt(crit \* p \* (1 - hh)/hh): NaNs produced

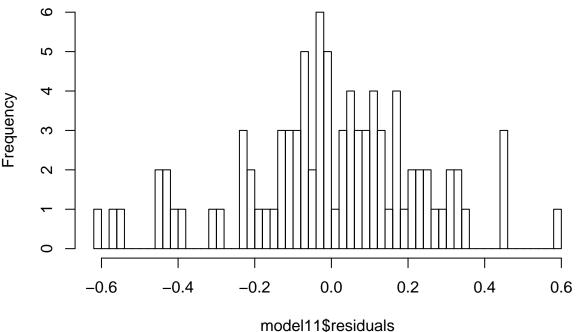
## Warning in sqrt(crit \* p \* (1 - hh)/hh): NaNs produced



Im(data\$crmrte\_1K\_log ~ prbarr + prbconv + pctmin80 + polpc\_1M\_log + wcon + ...

hist(model11\$residuals, breaks = 50)

## Histogram of model11\$residuals



```
coeftest(model11, vcov = vcovHC)
##
## t test of coefficients:
##
##
                  Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.0012e+00 7.5026e-01 -1.3345 0.1860116
## prbarr
                -2.2657e+00 2.7366e-01 -8.2790 3.216e-12 ***
## prbconv
                -7.0168e-01 1.1306e-01 -6.2061 2.632e-08 ***
## pctmin80
                1.1915e+00 1.8210e-01 6.5433 6.283e-09 ***
## polpc_1M_log 5.5062e-01 1.4578e-01 3.7771 0.0003132 ***
##
  wcon
                5.7789e-04
                            7.6194e-04
                                        0.7584 0.4505298
## wtuc
                6.1259e-05 4.9093e-04 0.1248 0.9010270
## wtrd
                1.4891e-03 1.2503e-03 1.1910 0.2373442
               -1.0503e-03 9.7864e-04 -1.0732 0.2865769
## wfir
## wmfg
                1.3441e-05 4.0958e-04
                                        0.0328 0.9739068
## wfed
                1.8267e-03 9.6458e-04 1.8938 0.0620559
## wsta
               -9.3939e-04 7.9649e-04 -1.1794 0.2419129
## wloc
               1.4813e-03 1.9299e-03 0.7675 0.4451429
##
  wser
               -1.4729e-04 1.0320e-03 -0.1427 0.8868930
##
  ---
## Signif. codes:
                  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#vcovHC(model11)
se.model11 = sqrt(diag(vcovHC(model11)))
stargazer(model10, model11,type="text",
se=list(se.model10, se.model11), star.cutoffs=c(0.05, 0.01, 0.001))
##
##
##
                                    Dependent variable:
```

(2)

crmrte\_1K\_log

(1)

## ##

##

```
## prbarr
                         -2.470***
                                              -2.266***
##
                          (0.269)
                                               (0.274)
##
## prbconv
                         -0.775***
                                               -0.702***
##
                          (0.105)
                                               (0.113)
##
## pctmin80
                          1.172***
                                            1.192***
##
                          (0.185)
                                               (0.182)
##
                          0.686***
                                              0.551***
## polpc_1M_log
##
                          (0.098)
                                                (0.146)
##
                                                 0.001
## wcon
                                                (0.001)
##
##
## wtuc
                                                0.0001
##
                                               (0.0005)
##
## wtrd
                                                 0.001
##
                                                (0.001)
##
                                                -0.001
## wfir
##
                                                (0.001)
##
                                                0.00001
## wmfg
##
                                               (0.0004)
##
## wfed
                                                 0.002
##
                                                (0.001)
##
## wsta
                                                -0.001
##
                                                (0.001)
##
## wloc
                                                 0.001
##
                                                (0.002)
##
                                                -0.0001
## wser
##
                                                (0.001)
##
## Constant
                          -0.822
                                                -1.001
                          (0.736)
##
                                                (0.750)
##
                           90
                                                90
## Observations
                          0.734
## R2
                                                0.812
## Adjusted R2
                       0.722
                                               0.780
## Residual Std. Error 0.290 (df = 85) 0.257 (df = 76)
## F Statistic 58.655*** (df = 4; 85) 25.253*** (df = 13; 76)
## Note:
                        *p<0.05; **p<0.01; ***p<0.001
linearHypothesis(model11, c("wcon = 0", "wtuc = 0", "wtrd = 0", "wfir = 0", "wmfg = 0", "wfed = 0", "wsta = 0"
## Linear hypothesis test
## Hypothesis:
## wcon = 0
## wtuc = 0
```

## wtrd = 0 ## wfir = 0

```
## wmfg = 0
## wfed = 0
## wsta = 0
## wloc = 0
##
## Model 1: restricted model
## Model 2: data$crmrte_1K_log ~ prbarr + prbconv + pctmin80 + polpc_1M_log +
       wcon + wtuc + wtrd + wfir + wmfg + wfed + wsta + wloc + wser
##
## Note: Coefficient covariance matrix supplied.
##
##
    Res.Df Df
                  F Pr(>F)
## 1
        84
## 2
        76 8 2.2669 0.0313 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```