## Untitled

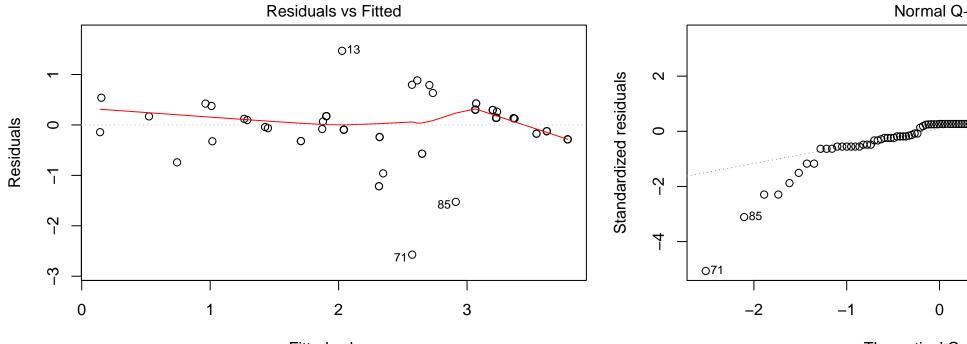
## Amanda McDermott

4/28/2019

```
[1] "c(\"100%\", \"33%\", \"97%\", \"69%\", \"99%\", \"99%\", \"100%\", \"75%\", \"100%\", \"98%\", \"93%\", \"100%\", \"99%\
## [2] "c(\"100%\", \"45%\", \"79%\", \"72%\", \"68%\", \"85%\", \"82%\", \"87%\", \"100%\", \"81%\", \"90%\", \"100%\", \"65%\"
## [3] "c(\"0%\", \"47%\", \"49%\", \"82%\", \"44%\", \"87%\", \"18%\", \"7%\", \"50%\", \"0%\", \"38%\", \"56%\", \"56%\", \"46%\", \"26%
       [4] "c(\"0%\", \"100%\", \"52%\", \"100%\", \"92%\", \"88%\", \"47%\", \"10%\", \"83%\", \"0%\", \"46%\", \"55%\", \"0%\", \"76%\", \"
        [5] "c(\"96%\", \"58%\", \"92%\", \"69%\", \"99%\", \"99%\", \"99%\", \"100%\", \"100%\", \"100%\", \"89%\", \"00%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\", \"100%\"
         [6] "c(\"57%\", \"53%\", \"52%\", \"67%\", \"72%\", \"28%\", \"71%\", \"24%\", \"67%\", \"73%\", \"59%\", \"93%\", \"57%\", \"98%\", \
        [7] "c(\"33%\", \"55%\", \"82%\", \"95%\", \"57%\", \"85%\", \"41%\", \"27%\", \"49%\", \"70%\", \"72%\", \"54%\", \"44%\", \"48%\", \
        [8] "c(\"100%\", \"100%\", \"95%\", \"94%\", \"93%\", \"90%\", \"89%\", \"87%\", \"87%\", \"77%\", \"77%\", \"72%\", \"71%\", \"71%\",
## [9] "c(\"0.39\", \"0.61\", \"0.69\", \"0.59\", \"0.66\", \"0.63\", \"0.56\", \"0.39\", \"0.61\", \"0.43\", \"0.57\", \"0.68\", \"0.35\
\#\# [10] "c(0.76, 0.25, 0.29, 0.39, 0.31, 0.24, 0.5, 0.59, 0.31, 0.7, 0.4, 0.23, 0.78, 0.19, 0.43, 0.77, 0.59, 0.56, 0.16, 0.74, 0.1, 0.13,
## [11] "c(0.33, 0.59, 0.52, 0.53, 0.58, 0.58, 0.4, 0.52, 0.55, 0.28, 0.43, 0.57, 0.37, 0.63, 0.27, 0.29, 0.43, 0.51, 0.65, 0.25, 0.64, 0.55
## [12] "c(0.77, 0.29, 0.53, 0.42, 0.37, 0.31, 0.62, 0.45, 0.35, 0.76, 0.52, 0.34, 0.76, 0.28, 0.73, 0.85, 0.66, 0.6, 0.26, 0.87, 0.19, 0.
## [13] "c(0.5, 0.49, 0.46, 0.52, 0.49, 0.46, 0.43, 0.44, 0.52, 0.46, 0.51, 0.54, 0.43, 0.55, 0.53, 0.45, 0.45, 0.48, 0.48, 0.47, 0.55, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 0.50, 
## [15] "c(\"0%\", \"96%\", \"42%\", \"97%\", \"90%\", \"34%\", \"39%\", \"13%\", \"78%\", \"0%\", \"35%\", \"46%\", \"0%\", \"67%\", \"22
## [16] "c(\"93%\", \"94%\", \"85%\", \"84%\", \"80%\", \"71%\", \"70%\", \"68%\", \"67%\", \"46%\", \"45%\", \"34%\", \"33%\", \"33%\", \
## Warning in .f(.x[[i]], ...): NAs introduced by coercion
temp_holc <- holc[,c(1,2,3,4,5,6,7,8,9, 14)]
temp <- pvia 2 %>%
    filter(!is.na(fatalities)) %>%
    count(state, fatalities) %>%
    group by(state) %>%
    mutate(assassinations = sum(n)) %>%
    dplyr::select(-fatalities, -n) %>%
    unique() %>%
    left_join(pvia_2, by = "state") %>%
    left_join(temp_holc, by = c("city_holc" = "city")) %>%
    mutate(year = as.integer(str_extract(date, "\\d{4}"))) %>%
    left_join(gtd_nkill_sum, by = c("year" = "iyear")) %>%
    transform(AMI = AMUI * 100,
```

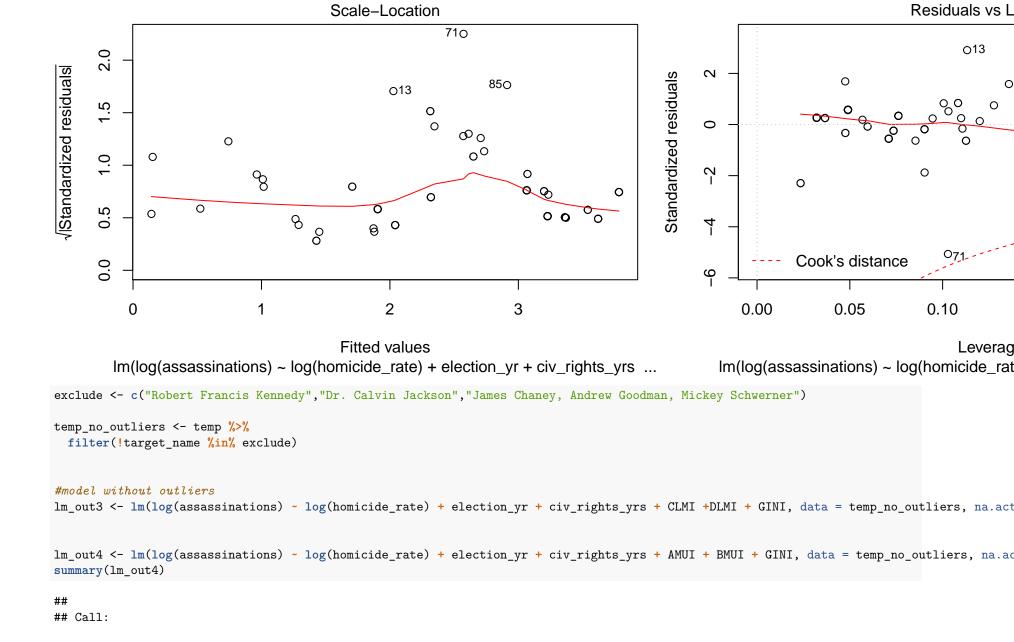
```
BMUI = BMUI * 100,
       CLMI = CLMI * 100,
       DLMI = DLMI * 100,
       GINI = GINI * 100)
lm_out <- lm(log(assassinations) ~ log(homicide_rate) +log(gdp_percapita_1980) + election_yr + civ_rights_yrs + gtd_incidents, data = temp</pre>
lm_out2 <- lm(log(assassinations) ~ log(homicide_rate) + election_yr + civ_rights_yrs + CLMI +DLMI + GINI, data = temp, na.action = na.omi</pre>
summary(lm_out2)
## Call:
## lm(formula = log(assassinations) ~ log(homicide_rate) + election_yr +
      civ_rights_yrs + CLMI + DLMI + GINI, data = temp, na.action = na.omit)
##
## Residuals:
      Min
              1Q Median
##
                            3Q
                                  Max
## -2.5729 -0.1732 0.1402 0.2633 1.4696
## Coefficients:
##
                    Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                    2.934403 1.331603 2.204 0.03050 *
## log(homicide_rate) 2.458465 0.258719 9.502 1.15e-14 ***
## election_yr
                   ## civ_rights_yrs
                   ## CLMI
                   -0.004266 0.005155 -0.828 0.41044
## DLMI
                   ## GINI
                   ## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.5363 on 78 degrees of freedom
   (78 observations deleted due to missingness)
## Multiple R-squared: 0.7539, Adjusted R-squared: 0.735
## F-statistic: 39.82 on 6 and 78 DF, p-value: < 2.2e-16
```



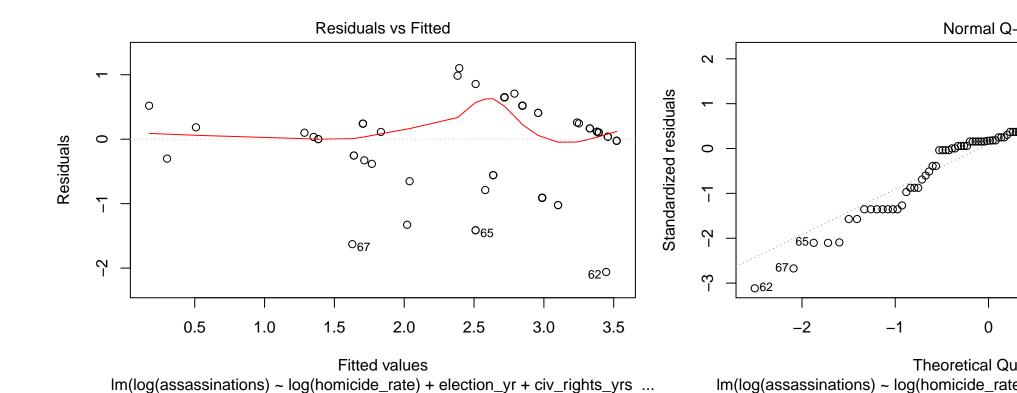


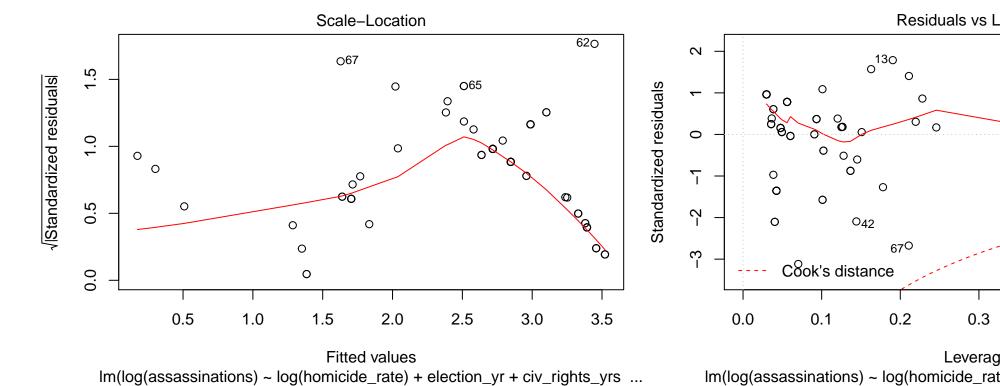
Fitted values Im(log(assassinations) ~ log(homicide\_rate) + election\_yr + civ\_rights\_yrs ...

Theoretical Qualin(log(assassinations) ~ log(homicide\_rate



```
## lm(formula = log(assassinations) ~ log(homicide_rate) + election_yr +
      civ_rights_yrs + AMUI + BMUI + GINI, data = temp_no_outliers,
##
      na.action = na.omit)
## Residuals:
      Min
              1Q Median
                            3Q
                                  Max
## -2.0600 -0.3214 0.1084 0.5206 1.1023
## Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                   -7.802042 1.631014 -4.784 8.42e-06 ***
## log(homicide_rate) 2.056177
                              0.346033 5.942 8.24e-08 ***
## election_yr
                    0.128914
                              0.170288 0.757 0.45140
                   ## civ_rights_yrs
## AMUI
                    2.489294 1.479527 1.682 0.09663 .
## BMUI
                    ## GINI
                    0.028258
                             0.028245
                                       1.000 0.32029
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6855 on 75 degrees of freedom
   (78 observations deleted due to missingness)
## Multiple R-squared: 0.5703, Adjusted R-squared: 0.536
## F-statistic: 16.59 on 6 and 75 DF, p-value: 4.495e-12
plot(lm_out4)
```





#stargazer(lm\_out, lm\_out2, lm\_out3, title = "Regression Results", order = c(1, 3, 2, 5, 4, 6, 7, 8), #covariate.labels = c("Log(1980 Homostargazer(lm\_out, lm\_out2, lm\_out3, title = "Regression Results", order = c(1,3,2,5,4,6,7,8), covariate.labels = c("Log(1980 Homostargazer(lm\_out, lm\_out2, lm\_out3, title = "Regression Results", order = c(1,3,2,5,4,6,7,8), covariate.labels = c("Log(1980 Homostargazer(lm\_out, lm\_out2, lm\_out3, title = "Regression Results", order = c(1,3,2,5,4,6,7,8), covariate.labels = c("Log(1980 Homostargazer(lm\_out, lm\_out2, lm\_out3, title = "Regression Results", order = c(1,3,2,5,4,6,7,8), covariate.labels = c("Log(1980 Homostargazer(lm\_out, lm\_out2, lm\_out3, title = "Regression Results", order = c(1,3,2,5,4,6,7,8), covariate.labels = c("Log(1980 Homostargazer(lm\_out, lm\_out2, lm\_out3, title = "Regression Results", order = c(1,3,2,5,4,6,7,8), covariate.labels = c("Log(1980 Homostargazer(lm\_out, lm\_out3, lm\_o

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu % Date and time: Mon, Apr 29, 2019 - 00:42:35

Table 1: Regression Results

		Inequality Model	Inequality Model w/o Outliers
	(1)	(2)	(3)
Log(1980 Homicide Rate)	1.191***	2.458***	2.544***
	(0.306)	(0.259)	(0.195)
Election Year Dummy	-0.171	-0.163	-0.014
	(0.199)	(0.128)	(0.100)
Log(1980 GDP Per Capita)	0.225		
	(0.417)		
Int'l Assassination Incidents	-0.0003		
	(0.0003)		
Civil Rights Year Dummy		-0.493**	-0.383**
		(0.203)	(0.168)
C Lower-Middle Income		-0.004	$-0.007^*$
		(0.005)	(0.004)
D Lower-Middle Income		-0.044***	-0.038***
		(0.008)	(0.006)
Gini Index		-0.063***	$-0.053^{***}$
		(0.024)	(0.018)
Constant	-2.689	2.934**	1.897*
	(3.780)	(1.332)	(1.046)
Observations	127	85	82
$\mathbb{R}^2$	0.199	0.754	0.852
Adjusted $R^2$	0.173	0.735	0.840
Residual Std. Error	0.995 (df = 122)	0.536 (df = 78)	0.402 (df = 75)
F Statistic	$7.579^{***} (df = 4; 122)$	$39.821^{***} (df = 6; 78)$	$72.061^{***} (df = 6; 75)$

Note: