

# 607 Final Project - Cause of Death by Firearm vs State Firearm Laws

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## Research question

Are there aspects of gun policy that are predictors of firearm mortality rate?

## Cases

All datasets and relevant information can be found in Github Repository - <https://github.com/JAbinette/CUNY-607-Final-Project>

Cause of Death by Firearm data retrieved from Centers for Disease Control & Prevention website includes 458 observations grouped by State, Year and Cause of Death Category (ICD Sub-Chapter) including firearm deaths excluding Terrorism, Legal Intervention and Operations of War See Github File: '0 - wonder.cdc.gov Underlying Cause of Death cdc - Grp by State, ICD Sub-Chapter.txt'

Firearm Laws by State data retrieved from <https://www.statefirearmlaws.org/resources> Database containing detailed annual information on firearm-related laws in place in each of the 50 US states includes 150 observations with 137 variables after data subset to only include 2018-2020. See Github File: '0 - state-firearmlaws.org Firearm Laws - DATABASE\_0.xlsx'

## Data Preparation

### CDC 2020 Summary of Death by Firearm

```
library(tidyr)
library(readxl)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
# Load CDC Cause of Death by Firearm text file from Github
txt.url = 'https://raw.githubusercontent.com/JAbinette/CUNY-607-Final-Project/main/0%20-%20wonder.cdc.g
cod.ld <- read.delim (txt.url, header=TRUE, sep = "\t")

cod.st_yr <- subset (cod.ld, Year.Code == 2020, select = c("State", "Year.Code", "Deaths", "Population")
  group_by(State, Year.Code) %>%
  mutate(Firearm_Deaths = sum(Deaths)) %>%
  mutate (Prop.of.Pop = (Firearm_Deaths/Population) ) %>%
  ungroup()
cod.st_yr <- subset (cod.st_yr, State != "District of Columbia", select =c("State", "Year.Code", "Firea
  distinct() %>%
  na.omit()
```

## CDC Reported Deaths by State in 2020

```
# Load CDC Deaths by State in 2020 text file from Github
txt.url = 'https://raw.githubusercontent.com/JAbinette/CUNY-607-Final-Project/main/0%20-%20wonder.cdc.g
cod.all <- read.delim (txt.url, header=TRUE, sep = "\t")
cod.all <- subset(cod.all, Year.Code == 2020 & State != "District of Columbia", select = c("State", "Dea

# Merge with Firearm Deaths data
COD <- merge( cod.st_yr, cod.all, by.x = "State", by.y = "State", all.x = TRUE) %>%
  mutate (Prop.of.Deaths = (Firearm_Deaths/Deaths) )
```

## State Firearm Laws (filtered to only include 2020)

```
# Download State Firearm Laws spreadsheet and save to your file directory (data saved in Github at http
# Set path to excel spreadsheet
path = "0 - statefirearmlaws.org Firearm Laws - DATABASE_0.xlsx"
laws_ld <- read_excel(path)
# Subset data to only include years 2018-2020
laws_ld <- subset (laws_ld, year == 2020)
# Transform from wide to long (excluding last column which is the total number of laws)
laws_ld2 <- pivot_longer ( laws_ld, cols = 3:136, names_to ="Variable", values_to = "Indicator" )
# Remove variable records where the law was not present for that year and State
laws_ld3 <- subset ( laws_ld2, Indicator == 1 )
head(laws_ld3)
```

```
## # A tibble: 6 x 5
##   state   year lawtotal Variable      Indicator
##   <chr>   <dbl>   <dbl> <chr>         <dbl>
## 1 Alabama 2020      10 invcommitment      1
## 2 Alabama 2020      10 danger              1
## 3 Alabama 2020      10 alcoholism          1
## 4 Alabama 2020      10 dealerh              1
## 5 Alabama 2020      10 permitconcealed      1
## 6 Alabama 2020      10 ccbackground          1
```

## Add Variable Categories to Merge with State Firearm

```
# Load Variable Category data for State Firearm Laws data
laws.cat_ld <- read.csv('https://raw.githubusercontent.com/JAbinette/CUNY-607-Final-Project/main/0%20-%')
# Replace spaces in Category with a period
laws.cat_ld$Category <- make.names (laws.cat_ld$Category)
# Merge
laws_ld4 <- merge( laws_ld3, laws.cat_ld, by.x = "Variable", by.y = "Variable.Name", all.x = TRUE)
laws_ld4 <- subset ( laws_ld4, select = c("state","year","lawtotal","Category"))

library(dplyr)
# Add count of laws by Category
laws_ld4 <- laws_ld4 %>%
  group_by(state, year, lawtotal, Category) %>%
  summarise(Category_count=n(),
            .groups = 'drop')
head(laws_ld4)
```

```
## # A tibble: 6 x 5
##   state   year lawtotal Category          Category_co-1
##   <chr>  <dbl>    <dbl> <chr>                <int>
## 1 Alabama 2020      10 Concealed.carry.permitting      4
## 2 Alabama 2020      10 Dealer.regulations            1
## 3 Alabama 2020      10 Domestic.violence            2
## 4 Alabama 2020      10 Prohibitions.for.high.risk.gun.possession  3
## 5 Alaska  2020       3 Buyer.regulations            1
## 6 Alaska  2020       3 Possession.regulations          1
## # ... with abbreviated variable name 1: Category_count
```

```
# Transform long to wide
laws <- pivot_wider( laws_ld4, names_from = "Category", values_from = "Category_count")
# Convert NA to zero
laws[is.na(laws)] <- 0

# Merge State Laws data with Cause of Death by State and Year
df <- merge (COD, laws, by.x = c("State"), by.y = c("state"), all.x = TRUE)
df <- subset(df, select = c("State", "Prop.of.Pop", "Prop.of.Deaths", "lawtotal"))
```

## Statistical Analyses

```
summary(df)
```

```
##      State      Prop.of.Pop      Prop.of.Deaths      lawtotal
## Length:50      Min.      :0.0000334      Min.      :0.00387      Min.      : 1.00
## Class :character 1st Qu.:0.0001093      1st Qu.:0.01078      1st Qu.:  9.25
## Mode  :character Median :0.0001390      Median :0.01360      Median : 20.50
##              Mean  :0.0001473      Mean  :0.01390      Mean   : 29.44
##              3rd Qu.:0.0001920      3rd Qu.:0.01736      3rd Qu.: 40.50
##              Max.  :0.0002744      Max.   :0.03095      Max.   :111.00
```

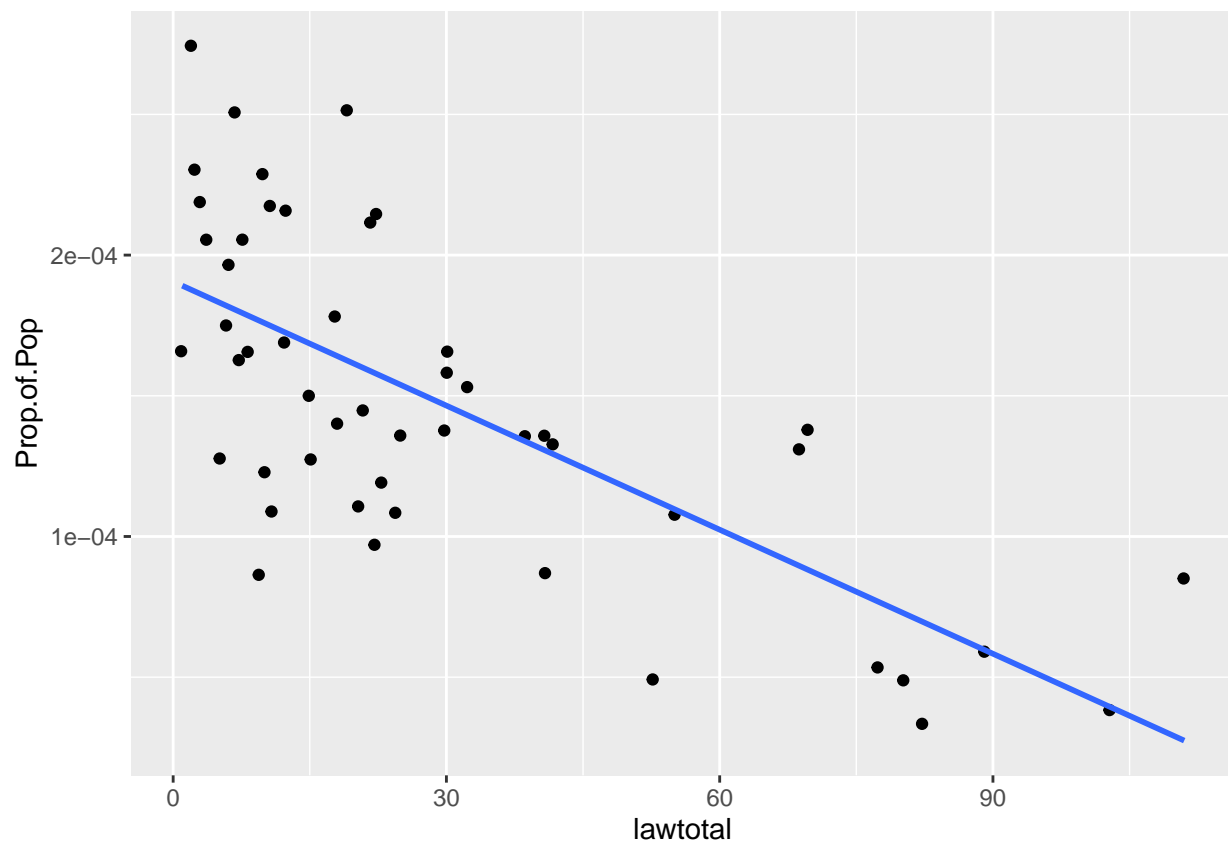
Can the Total Number of Laws predict proportion of Firearm deaths per State Population?

```
df %>%  
  summarise( cor( lawtotal, Prop.of.Pop, use = "complete.obs"))
```

```
##   cor(lawtotal, Prop.of.Pop, use = "complete.obs")  
## 1 -0.7024303
```

```
library(ggplot2)  
ggplot(data = df, aes(x = lawtotal, y = Prop.of.Pop)) + geom_jitter() + stat_smooth(method = "lm", se =
```

```
## 'geom_smooth()' using formula 'y ~ x'
```



```
# Simple Linear Regression  
m.pop <- lm(Prop.of.Pop ~ lawtotal, data = df)  
summary(m.pop)
```

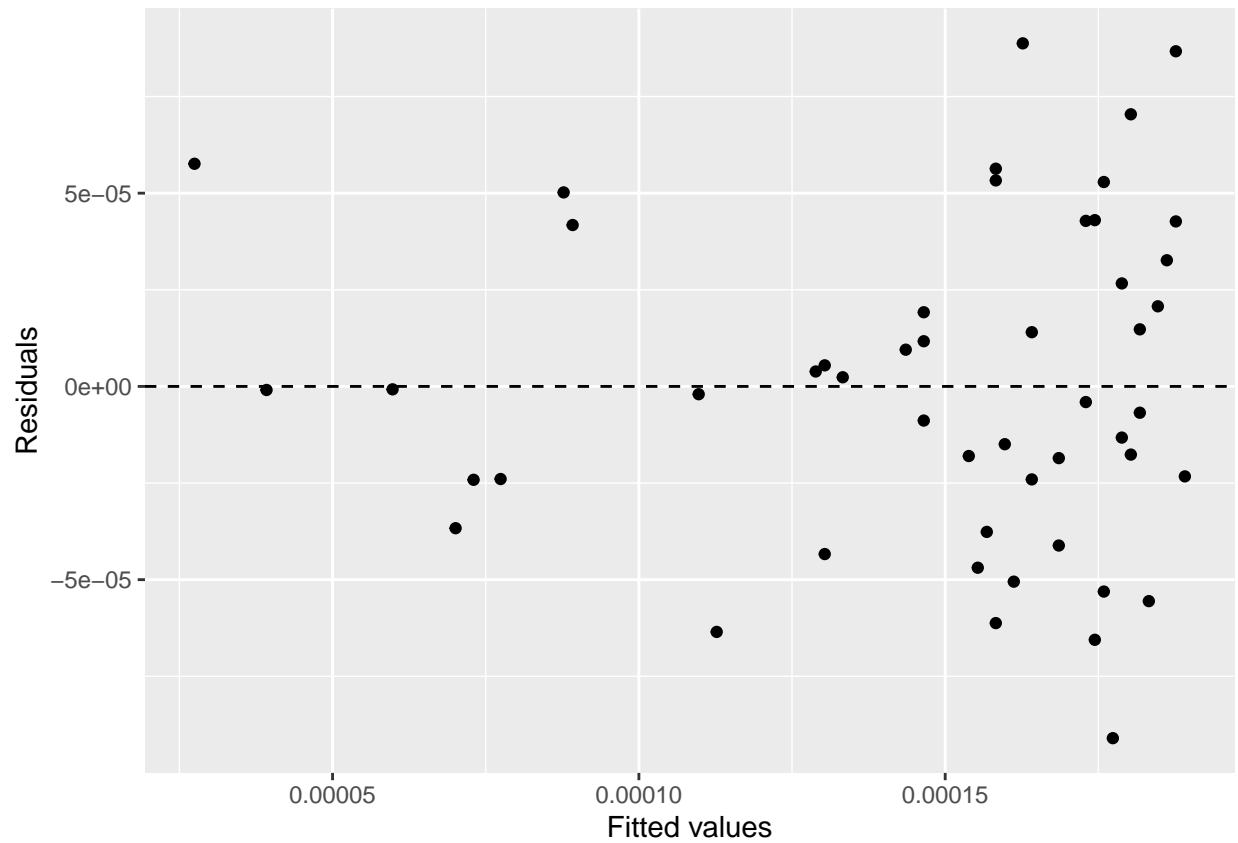
```
##  
## Call:  
## lm(formula = Prop.of.Pop ~ lawtotal, data = df)  
##  
## Residuals:
```

```
##           Min           1Q           Median           3Q           Max
## -9.100e-05 -2.413e-05 -1.467e-06  3.114e-05  8.876e-05
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.906e-04  8.727e-06  21.840  < 2e-16 ***
## lawtotal    -1.470e-06  2.149e-07  -6.837  1.31e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.249e-05 on 48 degrees of freedom
## Multiple R-squared:  0.4934, Adjusted R-squared:  0.4829
## F-statistic: 46.75 on 1 and 48 DF,  p-value: 1.305e-08
```

### Check Model Assumptions for Simple Regression

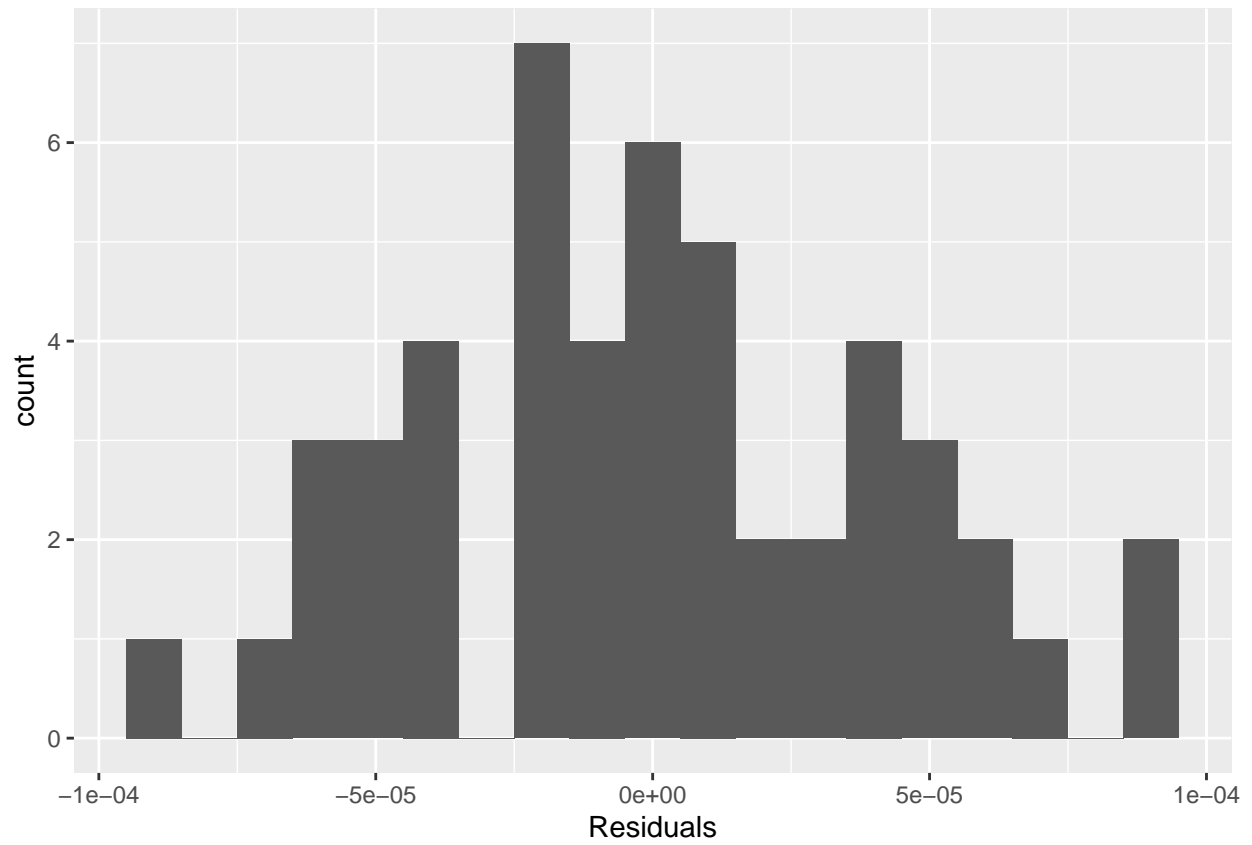
**Linearity and Constant Variability** - Conditions both met as there is no apparent pattern in the residuals plot indicating there is linearity and the points are scattered around zero showing constant variability.

```
ggplot(data = m.pop, aes(x = .fitted, y = .resid)) +
  geom_point() +
  geom_hline(yintercept = 0, linetype = "dashed") +
  xlab("Fitted values") +
  ylab("Residuals")
```

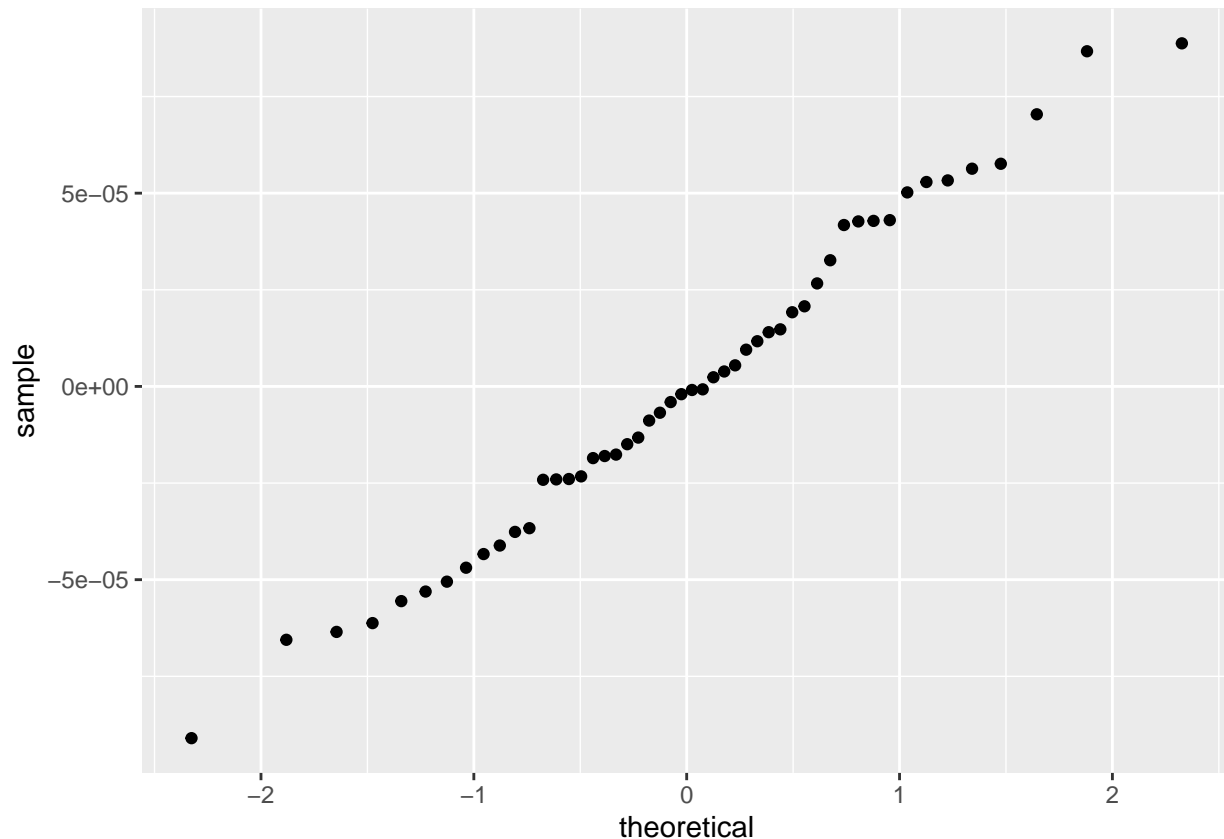


Nearly Normal Residuals - Condition is met based on below histogram and the normal probability plots.

```
ggplot(data = m.pop, aes(x = .resid)) +  
  geom_histogram(binwidth = .00001) +  
  xlab("Residuals")
```



```
ggplot(data = m.pop, aes(sample = .resid)) + stat_qq()
```



Can the Number and Category of Firearm Laws predict proportion of Firearm deaths per State Population?

```
# Merge State Laws data with Cause of Death by State and Year
df2 <- merge (COD, laws, by.x = c("State"), by.y = c("state"), all.x = TRUE)

# Multiple Linear Regression
m_2020 <- lm(Prop.of.Pop ~ Concealed.carry.permitting + Dealer.regulations + Domestic.violence +
Prohibitions.for.high.risk.gun.possession + Buyer.regulations + Possession.regulations + Stand.your.ground +
Background.checks + Child.access.prevention + Gun.trafficking + Immunity + Preemption, data = df2)
summary(m_2020)

##
## Call:
## lm(formula = Prop.of.Pop ~ Concealed.carry.permitting + Dealer.regulations +
##   Domestic.violence + Prohibitions.for.high.risk.gun.possession +
##   Buyer.regulations + Possession.regulations + Stand.your.ground +
##   Ammunition.regulations + Assault.weapons.and.large.capacity.magazines +
##   Background.checks + Child.access.prevention + Gun.trafficking +
##   Immunity + Preemption, data = df2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.912e-05 -1.866e-05 -4.747e-06  2.022e-05  9.049e-05
```



```
##
## Coefficients:
##
##               Estimate Std. Error t value
## (Intercept)      1.749e-04  1.387e-05  12.612
## Concealed.carry.permitting      1.298e-06  3.768e-06   0.344
## Dealer.regulations     -5.773e-07  3.243e-06  -0.178
## Domestic.violence     -1.357e-06  2.295e-06  -0.591
## Prohibitions.for.high.risk.gun.possession      1.717e-06  3.894e-06   0.441
## Buyer.regulations     -4.398e-06  3.330e-06  -1.321
## Possession.regulations      7.210e-06  4.265e-06   1.690
## Stand.your.ground     -3.039e-05  2.141e-05  -1.420
## Ammunition.regulations      1.221e-05  1.024e-05   1.192
## Assault.weapons.and.large.capacity.magazines  3.794e-06  5.031e-06   0.754
## Background.checks     -7.961e-07  2.845e-06  -0.280
## Child.access.prevention    -8.916e-06  3.997e-06  -2.231
## Gun.trafficking       -8.205e-06  7.430e-06  -1.104
## Immunity             -1.213e-06  2.057e-05  -0.059
## Preemption           -2.533e-05  1.217e-05  -2.082
##
##               Pr(>|t|)
## (Intercept)      1.41e-14 ***
## Concealed.carry.permitting      0.7326
## Dealer.regulations      0.8597
## Domestic.violence      0.5582
## Prohibitions.for.high.risk.gun.possession      0.6620
## Buyer.regulations      0.1951
## Possession.regulations      0.0998 .
## Stand.your.ground      0.1645
## Ammunition.regulations      0.2414
## Assault.weapons.and.large.capacity.magazines  0.4558
## Background.checks      0.7813
## Child.access.prevention      0.0322 *
## Gun.trafficking      0.2770
## Immunity      0.9533
## Preemption      0.0447 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.141e-05 on 35 degrees of freedom
## Multiple R-squared:  0.6493, Adjusted R-squared:  0.509
## F-statistic: 4.628 on 14 and 35 DF, p-value: 0.0001146
```

Fit the Best Model using backward-selection and p-value as the selection criterion

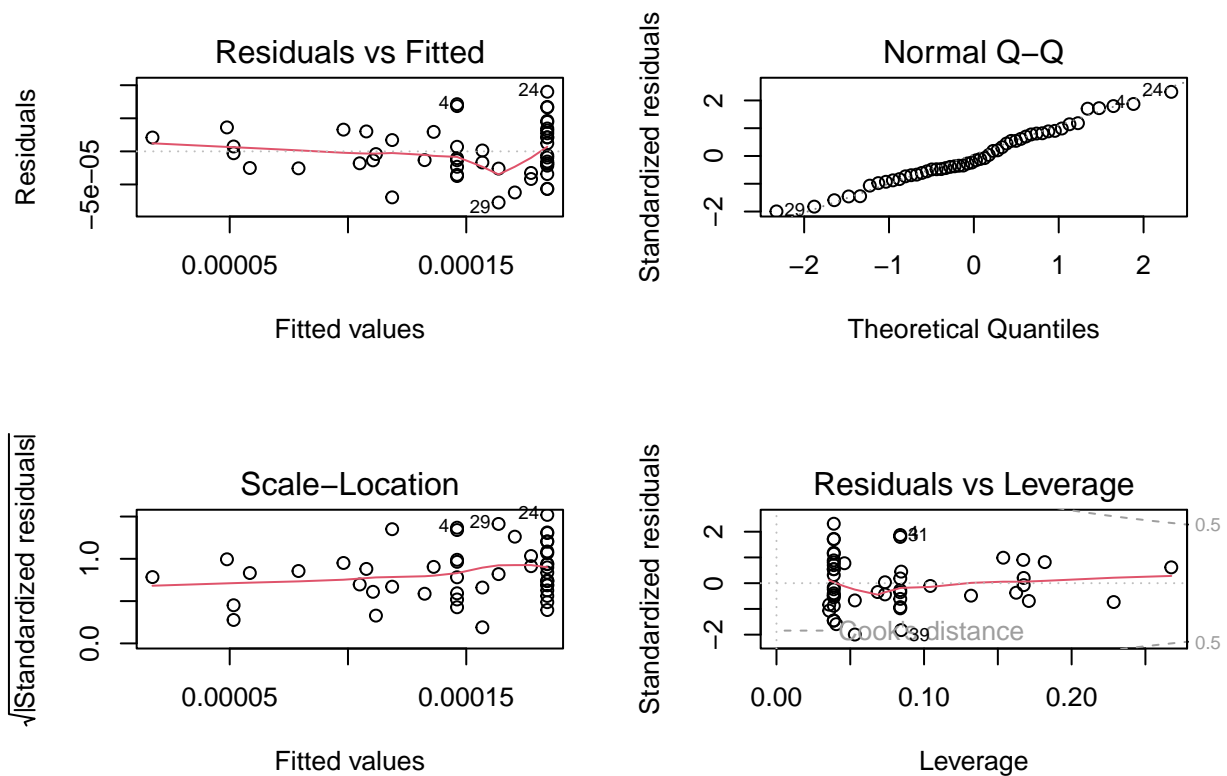
```
# Variables removed highest p-value at a time: Immunity, Dealer.regulations, Concealed.carry.permitting
m_bestfit <- lm(Prop.of.Pop ~ Stand.your.ground + Child.access.prevention + Preemption, data = df2)
summary(m_bestfit)

##
## Call:
## lm(formula = Prop.of.Pop ~ Stand.your.ground + Child.access.prevention +
##     Preemption, data = df2)
##
```

```
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.730e-05 -2.465e-05 -7.675e-06  2.893e-05  9.012e-05
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    1.842e-04  7.847e-06  23.480  < 2e-16 ***
## Stand.your.ground -3.813e-05  1.314e-05  -2.902  0.00567 **
## Child.access.prevention -6.860e-06  2.492e-06  -2.753  0.00843 **
## Preemption      -1.777e-05  7.425e-06  -2.393  0.02083 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.982e-05 on 46 degrees of freedom
## Multiple R-squared:  0.5737, Adjusted R-squared:  0.5459
## F-statistic: 20.64 on 3 and 46 DF,  p-value: 1.284e-08
```

### Check Model Assumptions

```
par(mfrow = c(2, 2))
plot(m_bestfit)
```



## **Conclusions**

The total number of firearm laws is a significant predictor and accounts for 49% of the variability in Firearm Deaths as a Proportion of the State Population. In breaking down the firearm laws by category, a multiple regression analysis determined that the number of Stand Your Ground, Child Access Prevention, and Preemption firearm laws can explain 55% of the variability.