# Bayes Project

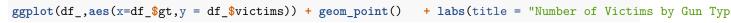
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#### Intro

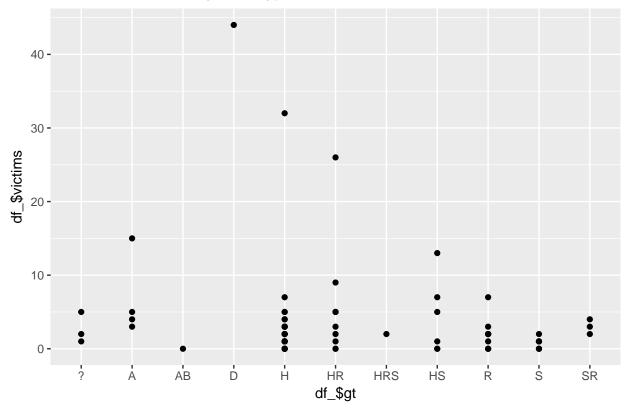
School shootings have long been unique in America for their frequency and intensity. We have obtained a dataset on all US school shootings since 1766 through UMass GRID in order to begin research on an issue that has had very little scientific study. By examining this data-set in a statistically rigorous way, we hope to gain a better understanding of a tragic phenomenon that does not seem to be in decline.

### Data

```
dates <- c("7/25/1880","3/30/1891","4/9/1891 ","3/26/1893","12/13/1898","1/17/1913","12/26/1916","5/18/
dates_formatted <- c()</pre>
for (i in 1:length(dates)){
 dates formatted <- c(dates formatted, as. Date(dates[i], "%m/%d/%Y"))
city <- c("Baton Rouge", "Liberty", "Newburgh", "Plain Dealing", "Charlestown", "Honolulu", "Lincoln ", "Bath"
county <- c("East Baton Rouge Parish", "Amite", "Orange", "Bossier Parish", "Poca", "Honolulu", "Casey", "Cli</pre>
state <- c("LA", "MS", "NY", "LA", "WV", "HI", "KY", "MI", "CA", "TX", "TX", "AZ", "OH", "WA", "IL", "NY", "CA", "I
victims <- c("1","0","0","2","5","1","1","44","5","5","15","5","4","1","1","3","1","7","2","2","2","0",
df_ <- data.frame(id = 1:84)</pre>
df_$dates <- as.factor(dates)</pre>
df_$dates <- factor(df_$dates , levels = dates)</pre>
df_$gt <- as.factor(gt)</pre>
df_$r <- r
df_$victims <- as.numeric(victims)</pre>
df_$state <- state</pre>
df_$count <- county</pre>
df_$city <- city</pre>
library(ggplot2)
```

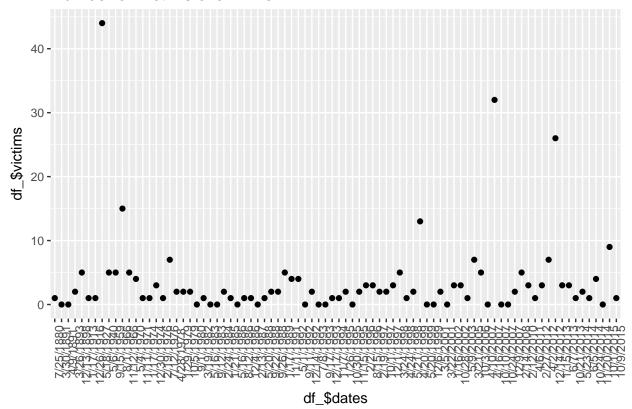


# Number of Victims by Gun Type



ggplot(df\_,aes(x=df\_\$dates,y = df\_\$victims)) + geom\_point() + labs(title = "Number of Victims over Time

## Number of Victims over Time



Potential model: Let

 $y_i =$  number of victims in incident  $r_i =$  race of the perpetrator

 $g_i = \text{gun used by perpetrator}$ 

 $s_i = \text{type of institution}$ 

 $\alpha_{i}[i] = \text{ random intercept accounting for geographic region}$ 

$$Y_i \sim Pois(\lambda_i)$$
 
$$log(\lambda_i) = \alpha_{j[i]} + \beta_1 * r_i + \beta_2 * g_i + \beta_3 * t_i + \beta_4 * s_i$$

### library(rjags)

## Loading required package: coda

## Linked to JAGS 4.3.0

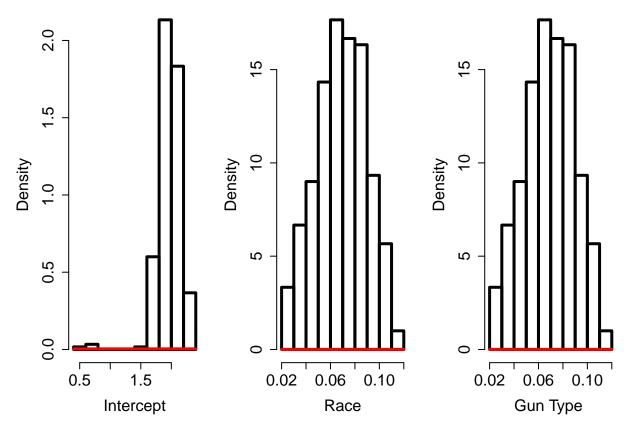
## Loaded modules: basemod, bugs

library(R2jags)

##

## Attaching package: 'R2jags'

```
## The following object is masked from 'package:coda':
##
##
       traceplot
model <- "
model {
for( i in 1:n) {
log(lambda.i[i]) <- (</pre>
# different notation; using betas for main effects and b's for varying effects
  + b1*r[i] + b2*gt[i]
 y.i[i] ~ dpois(lambda.i[i])
b0 \sim dnorm(0.0, 1.0E-4)
b1 \sim dnorm(0.0, 1.0E-4)
b2 \sim dnorm(0.0, 1.0E-4)
} # end model
jags.data <- list(y.i = df_$victims,</pre>
                   r = as.factor(df_$r), gt =as.factor(df_$gt), n=length(df_$victims))
parnames <- c( "b1", "b0", "b2")
mod <- jags(data = jags.data,
                    parameters.to.save=parnames,
                    n.chains = 3, n.burnin = 1500, n.iter = 1500 + 1000, n.thin = 10, model.file = textC
## module glm loaded
## Compiling model graph
##
      Resolving undeclared variables
##
      Allocating nodes
## Graph information:
      Observed stochastic nodes: 84
##
      Unobserved stochastic nodes: 3
##
##
      Total graph size: 338
##
## Initializing model
# some priors and posteriors to check that priors weren't informative
mcmc.array <- mod$BUGSoutput$sims.array</pre>
par(lwd = 3, cex.axis = 1.5, cex.lab = 1.5, cex.main = 1.5, mar = c(5,5,1,1), mfrow = c(1,3))
hist(c(mcmc.array[,,"b0"]), freq = F, main = "", xlab ="Intercept")
curve(dnorm(x,0,sqrt(1/10^(-4))), add = T, col = 2)
hist(c(mcmc.array[,,"b1"]), freq = F, main = "", xlab = "Race")
curve(dnorm(x,0,sqrt(1/10^{-4}))), add = T, col = 2)
hist(c(mcmc.array[,,"b1"]), freq = F, main = "", xlab = "Gun Type ")
curve(dnorm(x,0,sqrt(1/10^(-4))), add = T, col = 2)
```



We will also look at interactions and possible transformations of covariates as needed.

### Questions:

- Would a time series component work for this model?
- Should institution type be a hierarchical structure?