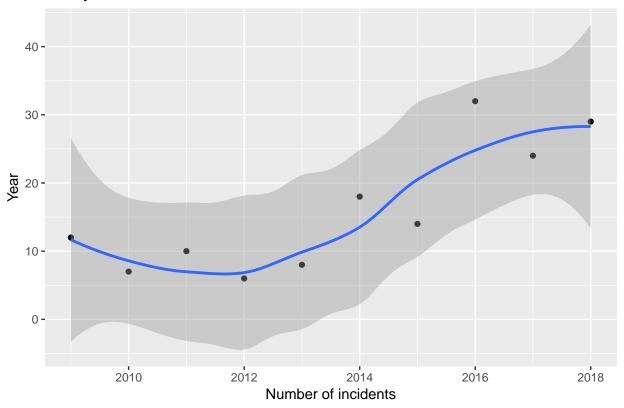
mass

SaiNagaChandraVivekGarimella August 16, 2019

```
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
library(tidyverse)
## -- Attaching packages ------
## v ggplot2 3.1.0
                    v readr
                              1.3.1
## v tibble 1.4.2
                    v purrr
                              0.2.5
## v tidyr 0.8.2 v stringr 1.3.1
## v ggplot2 3.1.0 v forcats 0.3.0
## -- Conflicts ------ tidyverse_confli
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
library(ggplot2)
library(lubridate)
## Attaching package: 'lubridate'
## The following object is masked from 'package:base':
##
##
      date
main <- read_csv('C:/Users/garim/OneDrive/Desktop/NEU/Introduction to Data Management and Processing/data
## Parsed with column specification:
## cols(
##
    LONCOD = col_double(),
##
    LATCOD = col_double(),
    CDCODE = col_double(),
##
##
    year = col_double(),
##
    school = col_character(),
##
    city = col_character(),
##
    state = col_character(),
##
    address = col_character(),
    urbanrural = col_character(),
##
##
    race = col_character(),
    killed = col_double(),
##
    injured = col_double(),
```

```
##
    victims = col_double(),
##
    date = col_character(),
##
    type = col_character(),
    NCESSCH = col_double(),
##
##
    time = col_time(format = "")
## )
main<-drop_na(main)</pre>
main
## # A tibble: 160 x 17
     LONCOD LATCOD CDCODE year school city state address urbanrural race
      <dbl> <dbl> <dbl> <dbl> <chr> <chr> <chr> <chr> <chr>
##
                                                                     <chr>
## 1 -75.6
                     1000 2009 Willi~ New ~ Dela~ 713 E ~ suburbtown BLACK
              39.7
## 2 -87.6
              41.8
                    1701 2009 Dunba~ Chic~ Illi~ 3000 S~ urban
## 3 -90.2
                    1712 2009 Cahok~ Caho~ Illi~ 800 Ra~ suburbtown BLACK
              38.6
   4 -78.3
              35.8 3713 2009 Zebul~ Zebu~ Nort~ Halifa~ suburbtown BLACK
##
## 5 -79.0
              35.1 3702 2009 Westo~ Faye~ Nort~ 277 Bo~ urban
                                                                     BLACK
              37.5 5103 2009 Chimb~ Rich~ Virg~ 3000 E~ urban
##
   6 -77.4
                                                                     BLACK
## 7 -84.5
                     2106 2009 Leest~ Lexi~ Kent~ 2010 L~ urban
              38.1
                                                                     BLACK
## 8 -92.9
              42.6
                    1901 2009 Aplin~ Park~ Iowa 610 N ~ rural
                                                                     WHITE
## 9 -72.5
                     3601 2009 Matti~ Matt~ New ~ 15125 ~ suburbtown WHITE
              41.0
## 10 -118.
              33.8
                     637 2009 Wilso~ Long~ Cali~ 4400 E~ urban
                                                                     HISP
## # ... with 150 more rows, and 7 more variables: killed <dbl>,
      injured <dbl>, victims <dbl>, date <chr>, type <chr>, NCESSCH <dbl>,
## #
      time <time>
#YEAR SMOOTH
main %>% group_by(year) %>% summarise(total=n()) %>% ggplot(mapping = aes(x=year,y=total))+geom_point()
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

Yearly Trend

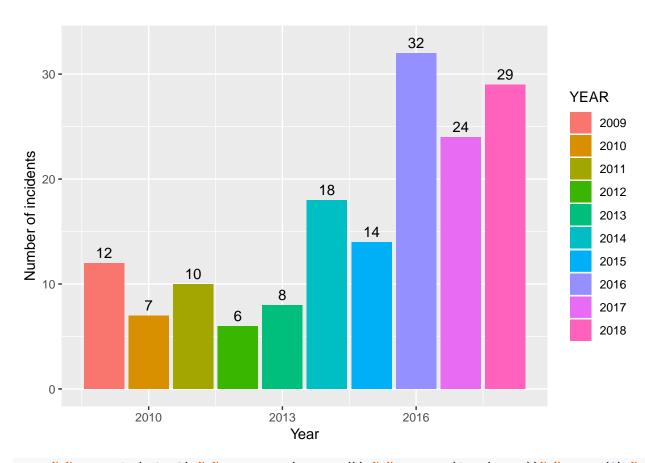


main

```
## # A tibble: 160 x 17
     LONCOD LATCOD CDCODE year school city state address urbanrural race
##
      <dbl>
             <dbl>
                   <dbl> <dbl> <chr> <chr> <chr> <chr>
                                                                   <chr>
##
   1
     -75.6
              39.7
                    1000 2009 Willi~ New ~ Dela~ 713 E ~ suburbtown BLACK
   2 -87.6
                    1701 2009 Dunba~ Chic~ Illi~ 3000 S~ urban
##
              41.8
                                                                   BLACK
##
   3 -90.2
              38.6
                    1712 2009 Cahok~ Caho~ Illi~ 800 Ra~ suburbtown BLACK
   4 -78.3
                    3713 2009 Zebul~ Zebu~ Nort~ Halifa~ suburbtown BLACK
              35.8
##
##
   5 -79.0
              35.1
                    3702 2009 Westo~ Faye~ Nort~ 277 Bo~ urban
                                                                   BLACK
##
   6 -77.4
              37.5
                    5103 2009 Chimb~ Rich~ Virg~ 3000 E~ urban
                                                                   BLACK
                    BLACK
##
   7 -84.5
              38.1
                    1901 2009 Aplin~ Park~ Iowa 610 N ~ rural
     -92.9
              42.6
##
                                                                   WHITE
   9 -72.5
              41.0
                    3601 2009 Matti~ Matt~ New ~ 15125 ~ suburbtown WHITE
##
## 10 -118.
              33.8
                     637 2009 Wilso~ Long~ Cali~ 4400 E~ urban
                                                                   HISP
## # ... with 150 more rows, and 7 more variables: killed <dbl>,
      injured <dbl>, victims <dbl>, date <chr>, type <chr>, NCESSCH <dbl>,
      time <time>
```

```
#YEAR BAR GRAPH
```

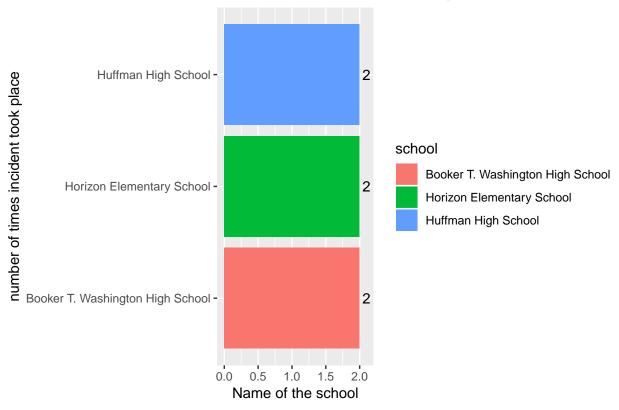
main %>% group_by(year) %>% summarise(count=n()) %>% ggplot(mapping = aes(x=year,y=count,fill=as.factor



main %>% group_by(school) %>% summarise(times=n()) %>% arrange(desc(times))%>% top_n(3) %>% ggplot(mapp

Selecting by times

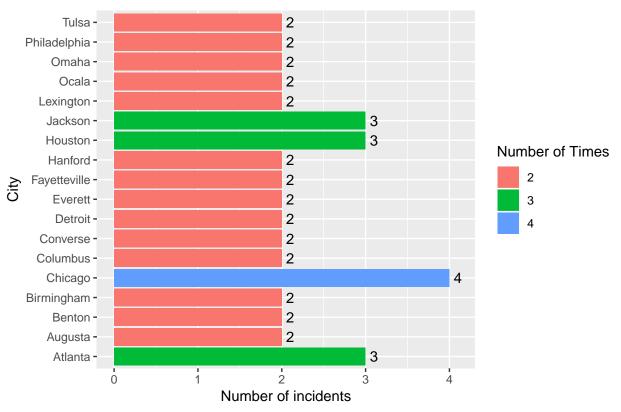
Schools where incidents took place more than once



#CITIES
main %>% group_by(city) %>% summarise(number_of_times=n()) %>% arrange(desc(number_of_times))%>% top_n(

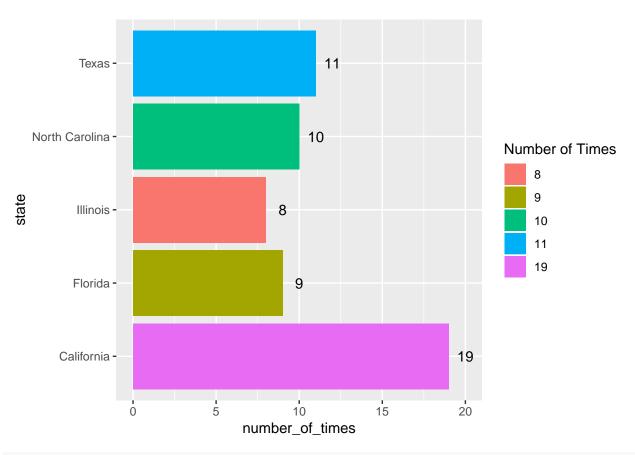
Selecting by number_of_times

Cities with most number of incidents



#STATES
main %>% group_by(state) %>% summarise(number_of_times=n()) %>% arrange(desc(number_of_times))%>% top_n

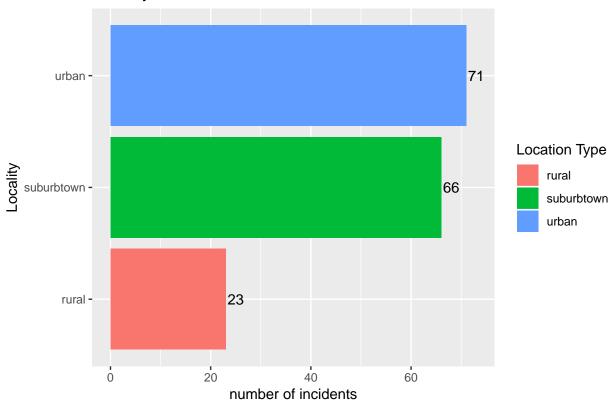
Selecting by number_of_times



#urbanrural
main %>% group_by(urbanrural) %>% summarise(number=n()) %>% arrange(desc(number))%>% top_n(5) %>% ggplo

Selecting by number

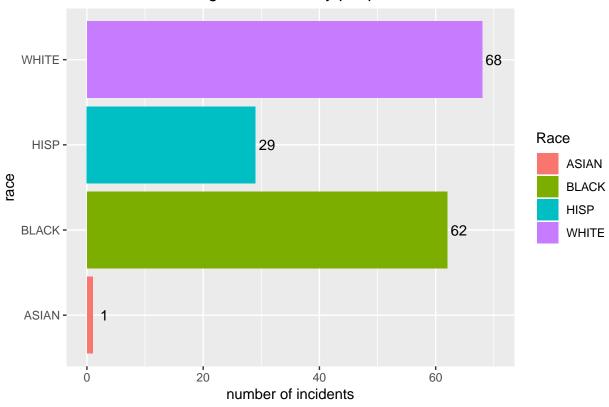
Locality vs Number of incidents



#Race
main %>% group_by(race) %>% summarise(counting=n()) %>% arrange(desc(counting))%>% top_n(5) %>% ggplot(s)

Selecting by counting

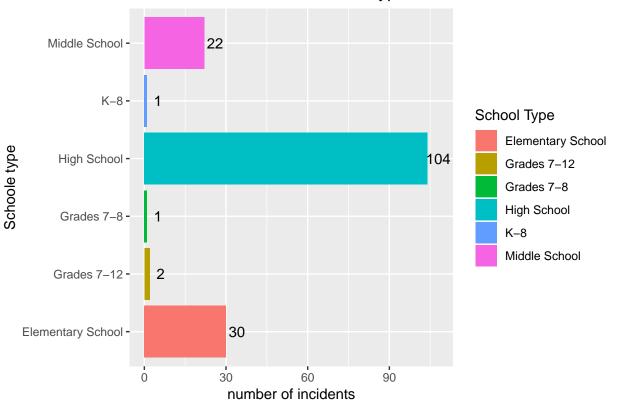




#SCHOOL TYPE
main %>% group_by(type) %>% summarise(total=n()) %>% arrange(desc(total))%>% top_n(5) %>% ggplot(mapping)

Selecting by total

Number of incidents in different types of Schools

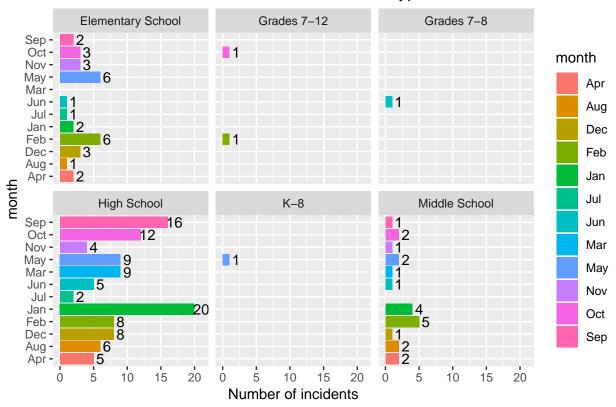


```
main<-main %>% mutate(month=str_sub(main[['date']],-6,-4))
main
```

```
## # A tibble: 160 x 18
     LONCOD LATCOD CDCODE year school city state address urbanrural race
##
##
      <dbl> <dbl> <dbl> <chr> <chr> <chr> <chr> <chr>
                                                         <chr>
                                                                   <chr>
##
   1 -75.6
              39.7
                    1000 2009 Willi~ New ~ Dela~ 713 E ~ suburbtown BLACK
   2 -87.6
                    1701 2009 Dunba~ Chic~ Illi~ 3000 S~ urban
##
              41.8
                                                                   BLACK
                    1712 2009 Cahok~ Caho~ Illi~ 800 Ra~ suburbtown BLACK
##
   3 -90.2
              38.6
##
   4 -78.3
              35.8
                    3713 2009 Zebul~ Zebu~ Nort~ Halifa~ suburbtown BLACK
##
   5 -79.0
              35.1
                    3702 2009 Westo~ Faye~ Nort~ 277 Bo~ urban
                                                                   BLACK
                    5103 2009 Chimb~ Rich~ Virg~ 3000 E~ urban
##
   6 -77.4
              37.5
                                                                   BLACK
##
      -84.5
              38.1
                    BLACK
                    1901 2009 Aplin~ Park~ Iowa 610 N ~ rural
   8 -92.9
              42.6
##
                                                                   WHITE
##
   9 -72.5
              41.0
                    3601 2009 Matti~ Matt~ New ~ 15125 ~ suburbtown WHITE
                     637 2009 Wilso~ Long~ Cali~ 4400 E~ urban
## 10 -118.
              33.8
                                                                   HISP
## # ... with 150 more rows, and 8 more variables: killed <dbl>,
      injured <dbl>, victims <dbl>, date <chr>, type <chr>, NCESSCH <dbl>,
      time <time>, month <chr>
#MONTHS EACH SCHOOL TYPE
main %>% group_by(month,type) %>% summarise(times_in_month=n()) %>% arrange(desc(times_in_month)) %>% t
```

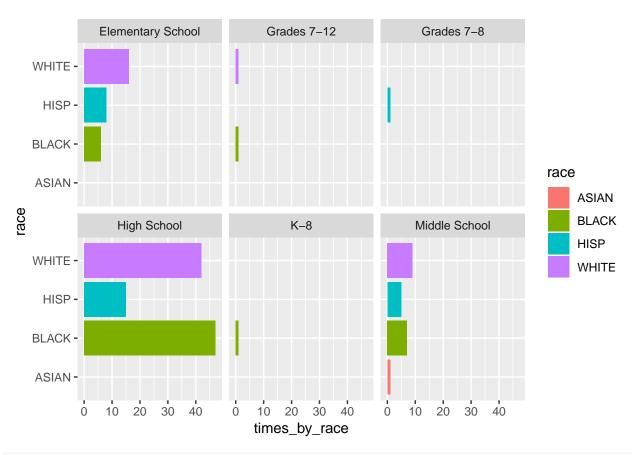
Selecting by times_in_month

Month vs Number of incident in each school type



#Race in each school type
main %>% group_by(race,type) %>% summarise(times_by_race=n()) %>% arrange(desc(times_by_race)) %>% top_:

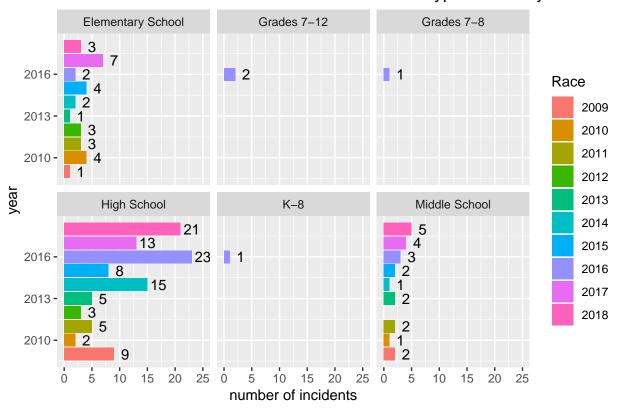
Selecting by times_by_race



#years in school type
main %>% group_by(year,type) %>% summarise(times_in_year=n()) %>% arrange(desc(times_in_year)) %>% top_:

Selecting by times_in_year

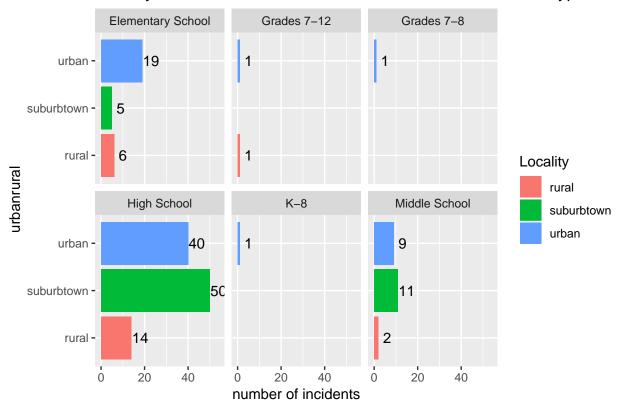
Number of incidents occurred at different school types in each year



urbanrural vs school type
main %>% group_by(urbanrural,type) %>% summarise(number=n()) %>% arrange(desc(number)) %>% top_n(5) %>%

Selecting by number

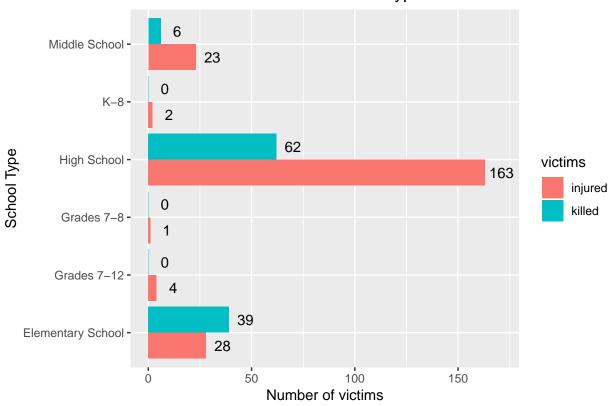
Locality vs Number of incidents occurred at different school types



#VICTIMS IN EACH SCHOOL TYPE
main %>% gather(key='victims',value='victims_count',killed,injured) %>% group_by(type,victims) %>% summ

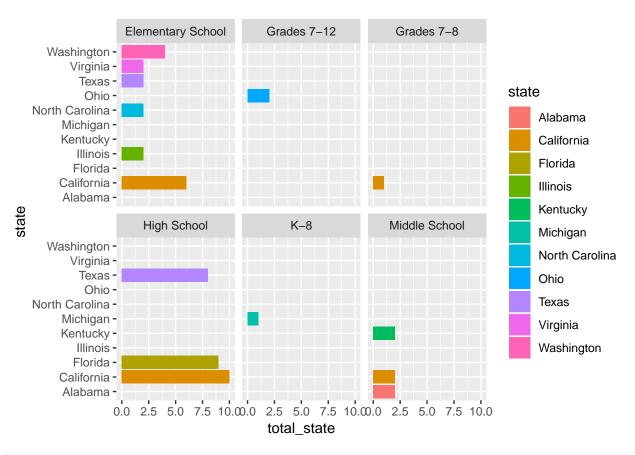
Selecting by total_victims

Number of victims in each school type



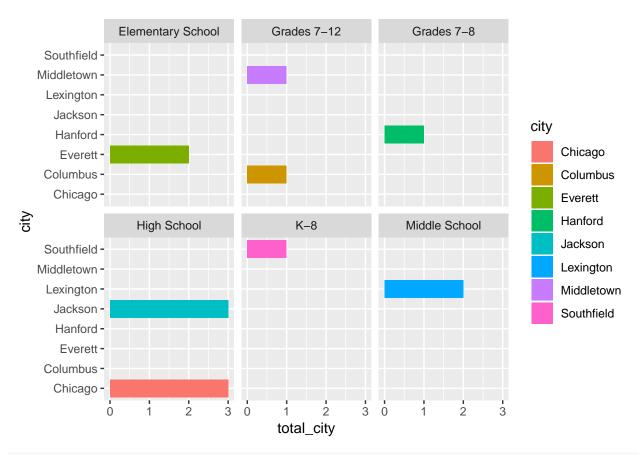
#state vs type
main %>% group_by(state,type) %>% summarise(total_state=n()) %>% arrange(desc(total_state)) %>% group_b

Selecting by total_state



#CITY VS TYPE
main %>% group_by(city,type) %>% summarise(total_city=n()) %>% arrange(desc(total_city)) %>% group_by(t

Selecting by total_city



one<-main %>% gather(key='victims',value='victims_count',killed,injured)