# NYPD Shooting Incident Data

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Analyzing police department shoot data provides us with a vast amount of information that can be used for good. Any time data analysis can be used for good then it affects us all. If shooting data is analysed optimally then communities can operate more smoothly and efficiently. They can do this because they can feel safe if the police departments know and understand the trends in their area. The most important of which is identifying specific locations where the incident rate is abnormally high and having the police department and the city officials develop ways to make those areas safer that doesn't involve stationing half of the police department there.

What better place to analyse then New York City. The diversity of both the population and the environment makes for an ideal data trove to be examined. Hopefully, we can discover elements that transcend jurisdictions and can help city officials and police departments in other states, or even countries.

#### Import Data

#### Tidy Data - Remove Unwanted Variables

### Tidy Data - Combining like catagories with Variable

```
clean_data <- clean_data %>%
  mutate(
    PERP_SEX = case_when(
    PERP_SEX %in% c("", "(null)", "U") ~ "Unknown",
    TRUE ~ as.character(PERP_SEX)
),
LOCATION_DESC = case_when(
    LOCATION_DESC %in% c("","(null)","0ther") ~ "0ther",
    TRUE ~ as.character(LOCATION_DESC)
),
LOC_CLASSFCTN_DESC = case_when(
    LOC_CLASSFCTN_DESC %in% c("","0THER","0ther") ~ "0ther",
    TRUE ~ as.character(LOC_CLASSFCTN_DESC)
),
```

```
PERP_AGE_GROUP = case_when(
    PERP_AGE_GROUP %in% c("","(null)","UNKNOWN") ~ "Unknown",
    TRUE ~ as.character(PERP_AGE_GROUP)
),
PERP_RACE = case_when(
    PERP_RACE %in% c("","(null)","UNKNOWN") ~ "Unkonwn",
    TRUE ~ as.character(PERP_RACE)
),
VIC_AGE_GROUP = case_when(
    VIC_AGE_GROUP %in% c("1022","UNKNOWN") ~ "Unknown",
    TRUE ~ as.character(VIC_AGE_GROUP)
),
LOC_OF_OCCUR_DESC = case_when(
    LOC_OF_OCCUR_DESC %in% c("") ~ "Unknown",
    TRUE ~ as.character(LOC_OF_OCCUR_DESC)
)
```

#### Tidy Data - Correct Data Types

```
library(lubridate)
clean_data$0CCUR_DATE <- mdy(clean_data$0CCUR_DATE)
clean_data$0CCUR_TIME <- hms(clean_data$0CCUR_TIME)
clean_data$BORO <- as.factor(clean_data$BORO)
clean_data$LOC_OF_OCCUR_DESC <- as.factor(clean_data$LOC_OF_OCCUR_DESC)
clean_data$LOC_CLASSFCTN_DESC <- as.factor(clean_data$LOC_CLASSFCTN_DESC)
clean_data$LOCATION_DESC <- as.factor(clean_data$LOCATION_DESC)
clean_data$PERP_AGE_GROUP <- as.factor(clean_data$PERP_AGE_GROUP)
clean_data$PERP_SEX <- as.factor(clean_data$PERP_SEX)
clean_data$PERP_RACE <- as.factor(clean_data$PERP_RACE)
clean_data$VIC_AGE_GROUP <- as.factor(clean_data$VIC_AGE_GROUP)
clean_data$VIC_SEX <- as.factor(clean_data$VIC_SEX)
clean_data$VIC_RACE <- as.factor(clean_data$VIC_RACE)
summary(clean_data)</pre>
```

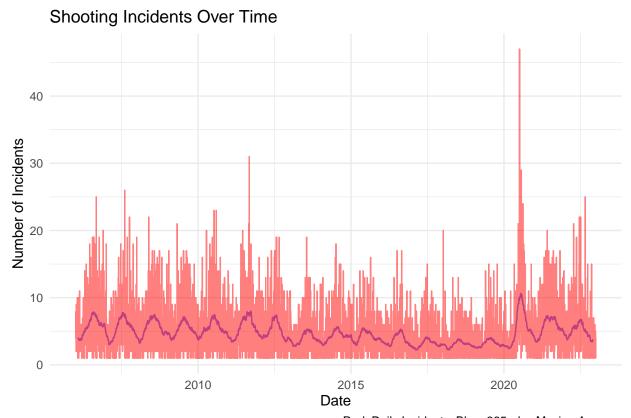
```
OCCUR_TIME
##
     OCCUR DATE
                                                                    BORO
## Min.
         :2006-01-01 Min.
                              :08
                                                         BRONX
                                                                     : 7937
  1st Qu.:2009-07-18 1st Qu.:3H 27M 0S
                                                         BROOKLYN
                                                                      :10933
## Median :2013-04-29 Median :15H 11M OS
                                                         MANHATTAN
                                                                      : 3572
## Mean :2014-01-06 Mean :12H 41M 31.7091388400731S
                                                         QUEENS
                                                                      : 4094
## 3rd Qu.:2018-10-15 3rd Qu.:20H 45M OS
                                                         STATEN ISLAND: 776
## Max. :2022-12-31 Max. :23H 59M OS
##
## LOC_OF_OCCUR_DESC LOC_CLASSFCTN_DESC
                                                        LOCATION_DESC
## INSIDE : 242
                    Other
                             :25627
                                       Other
                                                               :15954
## OUTSIDE: 1474
                              : 1103
                                       MULTI DWELL - PUBLIC HOUS: 4832
                    STREET
## Unknown:25596
                                       MULTI DWELL - APT BUILD : 2835
                    HOUSING : 280
                    DWELLING : 127
##
                                       PVT HOUSE
                                                               : 951
##
                    COMMERCIAL: 100
                                       GROCERY/BODEGA
                                                                  694
##
                    PLAYGROUND: 30
                                       BAR/NIGHT CLUB
                                                              : 628
##
                    (Other) : 45
                                       (Other)
                                                               : 1418
```

##	PERP_AGE_GROUP	PERP_SEX		PERP_RACE
##	Unknown:13132	F : 424	AMERICAN INDIAN/ALASKAN N	ATIVE: 2
##	18-24 : 6222	M :15439	ASIAN / PACIFIC ISLANDER	: 154
##	25-44 : 5687	Unknown:11449	BLACK	:11432
##	<18 : 1591		BLACK HISPANIC	: 1314
##	45-64 : 617		Unkonwn	:11786
##	65+ : 60		WHITE	: 283
##	(Other): 3		WHITE HISPANIC	: 2341
##	VIC_AGE_GROUP	VIC_SEX	VIC_R	ACE
##	<18 : 2839	F: 2615 AMERI	CAN INDIAN/ALASKAN NATIVE:	10
##	18-24 :10086	M:24686 ASIAN	/ PACIFIC ISLANDER :	404
##	25-44 :12281	U: 11 BLACK	:	19439
##	45-64 : 1863	BLACK	HISPANIC :	2646
##	65+ : 181	UNKNO	NN :	66
##	Unknown: 62	WHITE	:	698
##		WHITE	HISPANIC :	4049

### Visualizing the Data

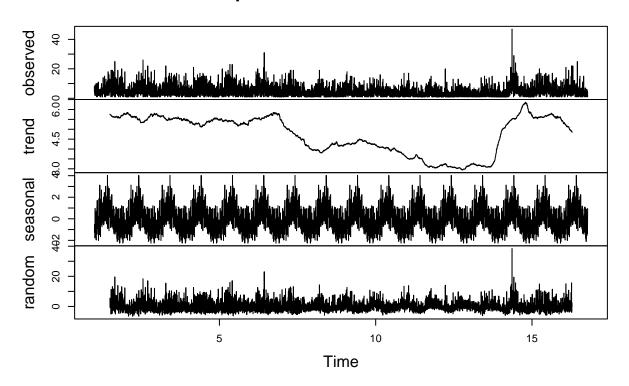
### Annual Trend related to number of Daylight Hours

## Warning: Removed 59 rows containing missing values ('geom\_line()').

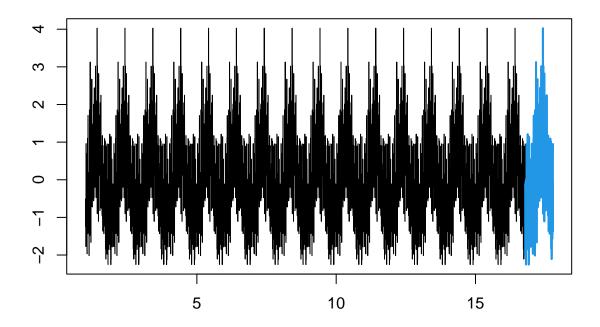


Red: Daily Incidents, Blue: 365-day Moving Average

# **Decomposition of additive time series**

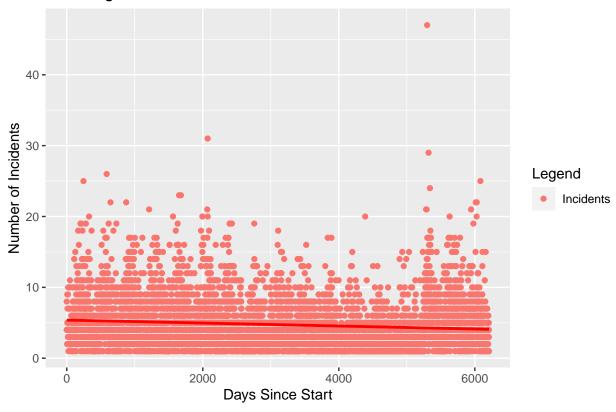


# Forecasts from STL + ETS(A,N,N)



```
##
## Call:
## lm(formula = incidents ~ day_num, data = incidents_by_day)
## Residuals:
##
     Min
             1Q Median
                           3Q
## -4.378 -2.500 -0.965 1.684 42.725
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 5.383e+00 9.242e-02 58.244 < 2e-16 ***
              -2.091e-04 2.592e-05 -8.064 8.88e-16 ***
## day_num
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 3.559 on 5759 degrees of freedom
## Multiple R-squared: 0.01117, Adjusted R-squared: 0.01099
## F-statistic: 65.03 on 1 and 5759 DF, p-value: 8.884e-16
## 'geom_smooth()' using formula = 'y ~ x'
```

### **Shooting Incidents Over Time**



#### Conclusion

There is evidence of seasonality in the rate of shooting incidents throughout the year. However, the reason for the trend is yet to be determined. One could argue that there are more hours of daylight in the summer months which means more people are out later than other times of year. This thesis could be tested by analyzing the time of day for each incident but I would believe that most incidents occur in the cover of night. Some bias that I'm experiencing with this data is that I would prefer that the incidents are more nightly than in the daytime because I am someone who is rarely out at night and would thus prefer the daytime to be a safer portion of the day. You could also argue that the hotter temperatures cause more irritability in people which leads to more aggressive encounters. This theory would be difficult to test without employing someone with a psychology background. Ultimately, there is plenty of data here to comb through and find meaningful correlations. The seasonality of the incidents is only the tip of the iceberg.