



DAT (DATA ANALYTICS TEAM) PRESENTS:

Predicting Crime in Queens.



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WHO ARE WE?

The background image shows a DJ in a nightclub. A vinyl record is spinning on a turntable in the foreground. In the background, a person is visible, holding a white cup. The scene is dimly lit with purple and blue ambient lighting.

OUR OBJECTIVE:

Using a variety of predictive modeling techniques, to what degree of accuracy can we predict the level of crime that will be committed in the bars and nightclubs of Queens, NY?

OUR DATASET:



NEW YORK CITY CRIMES
2014-2015.



PROBLEM FACED:

OVER 1 MILLION CRIME RECORDS

Dimensionality reduction

DATA CLEANING/PRE-PREPROCESSING:

```
nycrime_data <- read.csv(filepath)
nycrime_data <- nycrime_data[nycrime_data$BORO_NM == 'QUEENS',]
nycrime_data <- nycrime_data[nycrime_data$PREM_TYP_DESC == 'BAR/NIGHT CLUB',]
nycrime_data <- na.omit(nycrime_data)

write.csv(nycrime_data, 'clean_NYPD_Data.csv')
nycrime_data <- read.csv('clean_NYPD_Data.csv')

num <- c(3,4,9,12,13,16,17)

nycrime_data.clean <- subset(nycrime_data, select = c(num))
colnames(nycrime_data.clean) <- c('Date', 'Time', 'Desc', 'Success', 'LVO', 'Precinct', 'Occurrence')
nycrime_data.clean$Precinct <- as.factor(nycrime_data.clean$Precinct)
nycrime_data.clean$Date <- as.Date(nycrime_data.clean$Date, "%m/%d/%Y")

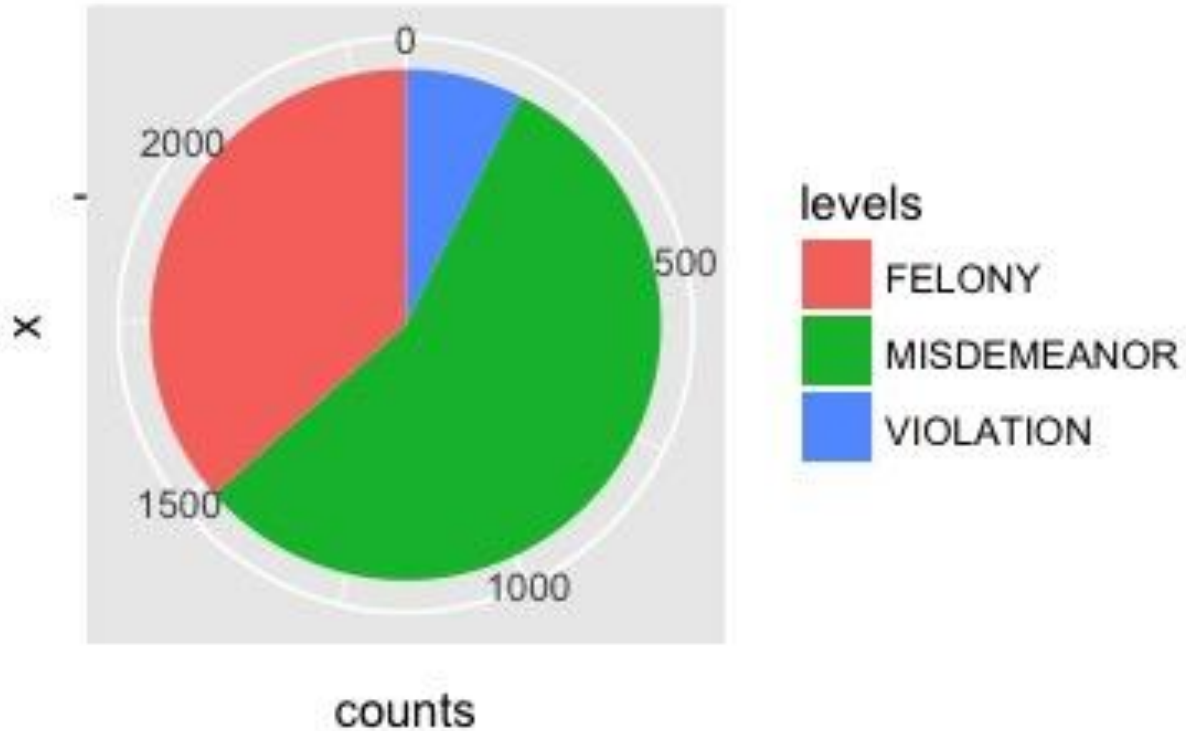
nycrime_data.clean <- nycrime_data.clean %>%
mutate(month = format(Date, "%m"), year = format(Date, "%Y")) %>%
group_by(month, year)
nycrime_data.clean <- nycrime_data.clean[,-1]

z <- lapply(str_split(nycrime_data.clean$Time, ":"), '[', 1)
nycrime_data.clean$Hour <- as.character(z)
nycrime_data.clean <- nycrime_data.clean[,-1]
cols <- c('month', 'year', 'Hour')
nycrime_data.clean[cols] <- lapply(nycrime_data.clean[cols], factor)
```

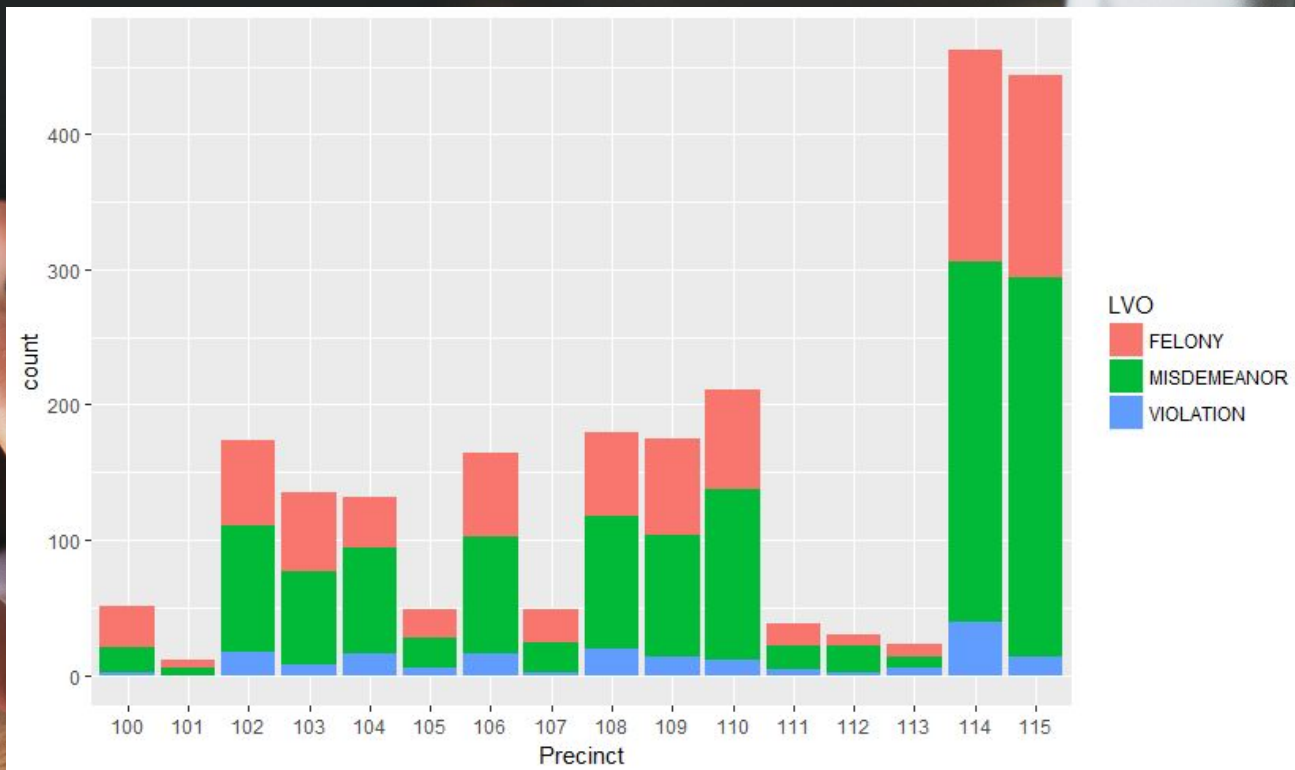

An overhead view of three people in an office setting, focused on reviewing architectural plans spread across a large wooden table. A man in a light blue shirt is on the left, pointing at a plan. A woman in a blue top is on the right, looking at the plans. A man in a light green checkered shirt is at the bottom, using a calculator. The table is cluttered with various items: a blue hard hat, a purple mug, a smartphone, a laptop, a ruler, and several rolled-up blueprints. The floor is covered in a dark, patterned carpet. A black banner with white text is centered over the image.

DESCRIPTIVE STATISTICS

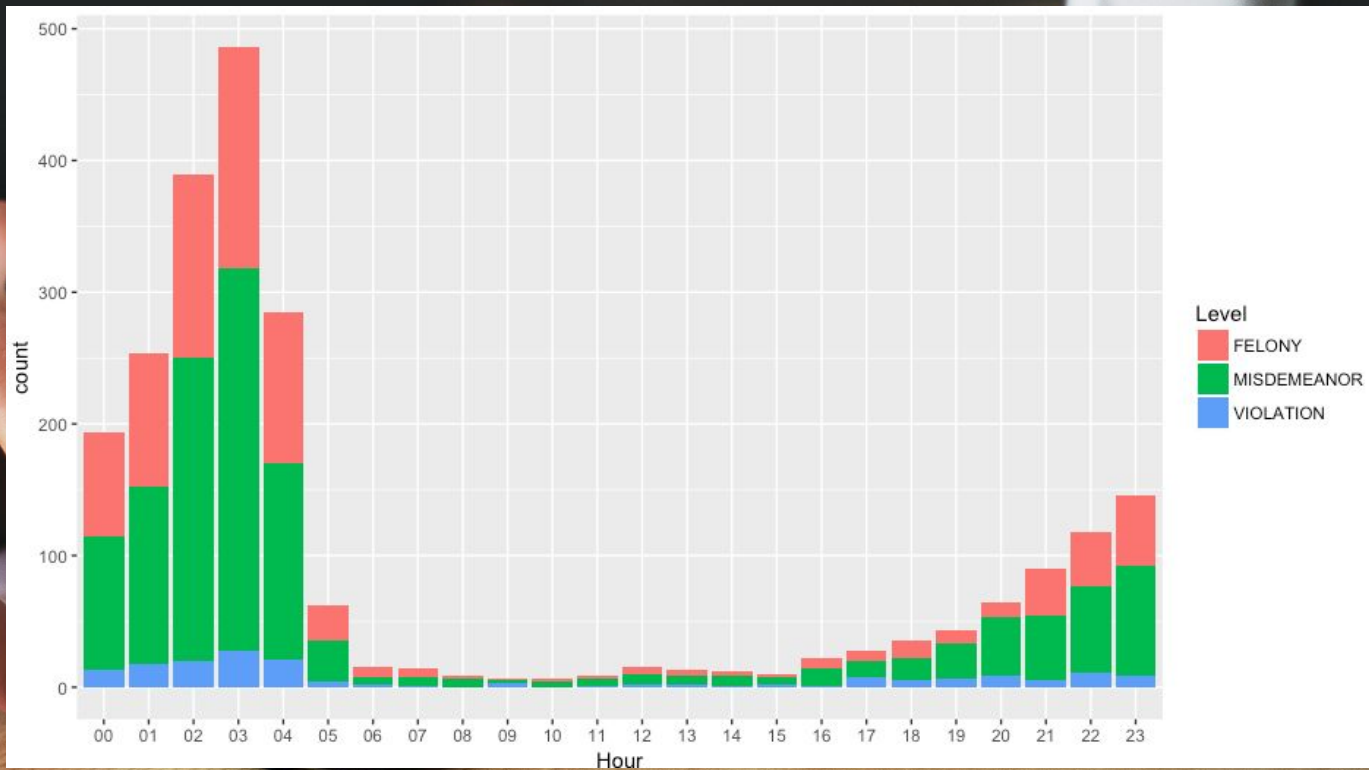
DESCRIPTIVE STATISTICS:



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OUR PLAN:

CMPLNT_FR_DT	Exact date of occurrence for the reported event (or starting date of occurrence, if CMPLNT_TO_DT exists)
CMPLNT_FR_TM	Exact time of occurrence for the reported event (or starting time of occurrence, if CMPLNT_TO_TM exists)
OFNS_DESC	Description of offense corresponding with key code
CRM_ATPT_CPTD_CD	Indicator of whether crime was successfully completed or attempted, but failed or was interrupted prematurely
LAW_CAT_CD	Level of offense: felony, misdemeanor, violation
ADDR_PCT_CD	The precinct in which the incident occurred
LOC_OF_OCCUR_DESC	Specific location of occurrence in or around the premises; inside, opposite of, front of, rear of

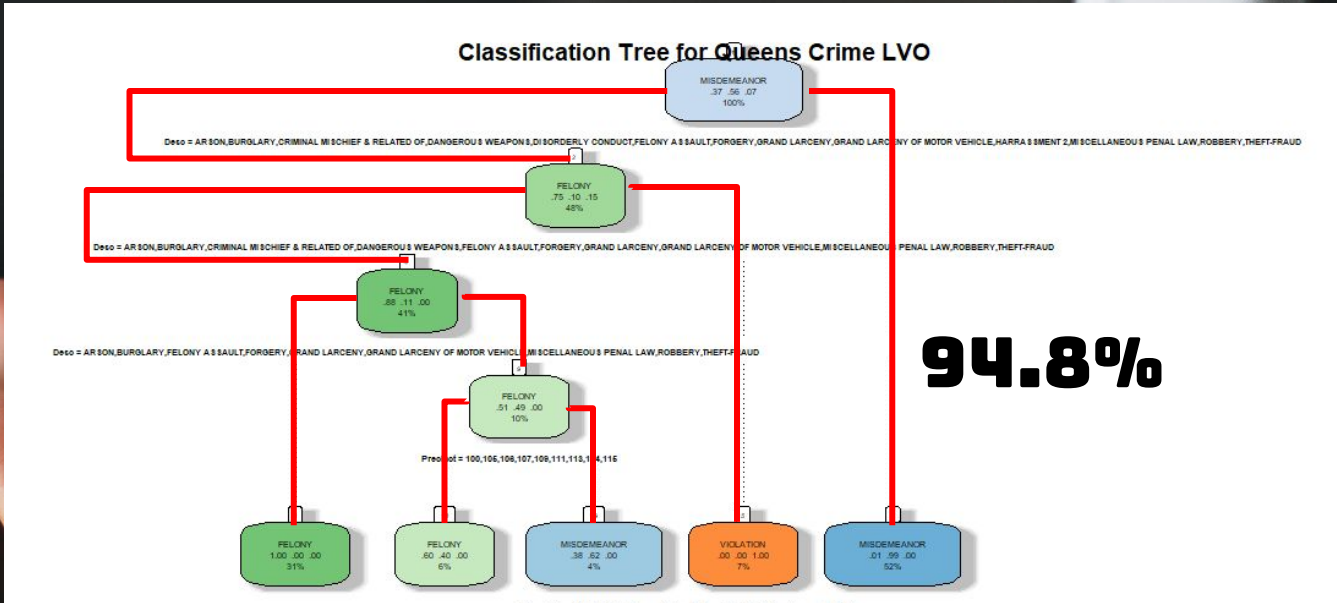
PREDICTIVE MODELING

- Decision Tree (RPart)
- Naive Bayes
- Support Vector Machine

DECISION TREE:

```
nyccrime.dtmodel <- rpart(LVO ~ ., data = nyccrime_data.clean,  
  method='class',  
  minsplit = 1,  
  minbucket = 5)
```


OUR RESULTS:



	Reference		
Prediction	FELONY	MISDEMEANOR	VIOLATION
FELONY	164	12	0
MISDEMEANOR	12	235	0
VIOLATION	0	0	43

MAIN SPLITTING ATTRIBUTES

- Crime Description
- Precinct

- Crime Description
- Precinct

NAIVE BAYES:

Initial

```
nycrime.nb <- train(x = train_x, y = train_y, method = 'nb', trControl=trainControl(method='cv', number=10))
nycrime.nb

nycrime.nb.predict <- predict(nycrime.nb$finalModel, test_x)

confusiontable.nb <- table(test_y, nycrime.nb.predict$class, dnn=c('Actual', 'Prediction'))
accuracy_initial <- sum(diag(confusiontable.nb)) / sum(confusiontable.nb)
accuracy_initial
```

DT Predictors

```
x <- nycrime_data.clean[,c(1,4)]
train_x <- x[index,]
test_x <- x[-index,]


nycrime.nb.updated <- train(x = train_x, y = train_y, method = 'nb', trControl=trainControl(method='cv', number=10))
nycrime.nb.updated

nycrime.nb.updated.predict <- predict(nycrime.nb.updated$finalModel, test_x)

confusiontable.nb.updated <- table(test_y, nycrime.nb.updated.predict$class, dnn=c('Actual', 'Prediction'))
accuracy_updated <- sum(diag(confusiontable.nb.updated)) / sum(confusiontable.nb.updated)
accuracy_updated
```


OUR RESULTS:

Initial



Actual	Prediction		
	FELONY	MISDEMEANOR	VIOLATION
FELONY	161	18	0
MISDEMEANOR	7	250	0
VIOLATION	0	0	30

DT Predictors

Actual	Prediction		
	FELONY	MISDEMEANOR	VIOLATION
FELONY	161	18	0
MISDEMEANOR	8	249	0
VIOLATION	0	0	30

SUPPORT VECTOR MACHINES:

Initial


```
train_data <- nycrime_data.clean[index,]  
test_data <- nycrime_data.clean[-index,]  
  
nycrime.svm <- svm(LVO~., data=train_data)  
nycrime.svm.predict <- predict(nycrime.svm, test_data)  
mean(nycrime.svm.predict == test_data$LVO)
```

DT Predictors

```
nycrime.svm.dt <- svm(LVO~Desc+Precinct, data=train_data)  
nycrime.svm.dt.predict <- predict(nycrime.svm.dt, test_data)  
mean(nycrime.svm.dt.predict == test_data$LVO)
```


OUR RESULTS:

Initial



Prediction	Reference		
	FELONY	MISDEMEANOR	VIOLATION
FELONY	172	21	0
MISDEMEANOR	7	236	0
VIOLATION	0	0	30

DT Predictors

Prediction	Reference		
	FELONY	MISDEMEANOR	VIOLATION
FELONY	149	0	0
MISDEMEANOR	30	257	0
VIOLATION	0	0	30

CONCLUSIONS:

- **ALL MODELS RELATIVELY ROBUST**
 - Achieved >93% for each algorithm used
 - With exception of decision tree, all built in very little time
- **EITHER LITTLE OR NO DIFFERENCE BETWEEN MODEL ITERATIONS**
- **OBSERVED OR INFERRED BIASES**
 - Crime is much more prevalent in certain precincts over others
 - May be at the behest of the opinion of whoever was the first responder
 - Majority of crime committed between the hours of 1-3 AM

QUESTIONS?

