NYU

The Anatomy of NYPD Crime Records:

Race, Gender, Politics and More

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Goals

- Intergroup Solidarity, Felt and Shared Foreignness between Asians and Hispanics in the U.S ✅
- They are under similar social pressure ?
  Similar crime behavior patterns in NYC (correlation based on crime amounts and types distribution) ❌

- **No hypothesis.**
- **All Races**
- **Various Analysis**
- **Direct Explaining**

Data

Unbiased data = How?

- Not arrest data but Complaint Data
- Omit minor crimes such as marijuwana, loitering, minor valued theft

Cleaned dataset = what columns?

- Race, Gender, Age, Crime type, Victim, Location.....

Methodologies
&
Corresponding Results

PART 02

Clustering Analysis

Smallest Cluster for races

OFNS_DESC	day	0.0	1.0	2.0	3.0	4.0	5.0	6.0	labelk
HARRASSMENT 2		1576.0	1613.0	1578.0	1589.0	1604.0	1676.0	1638.0	1

Hispanic

OFNS_DESC	day	0.0	1.0	2.0	3.0	4.0	5.0	6.0	labelk
HARRASSMENT 2		397.0	377.0	370.0	385.0	422.0	391.0	424.0	1

Asian

OFNS_DESC	day	0.0	1.0	2.0	3.0	4.0	5.0	6.0	labelk
HARRASSMENT 2		934.0	906.0	970.0	901.0	905.0	880.0	851.0	1

White

OFNS_DESC	day	0.0	1.0	2.0	3.0	4.0	5.0	6.0	labelk
HARRASSMENT 2		2782.0	2694.0	2759.0	2802.0	2761.0	2667.0	2637.0	1

Black

Clustering using K-means with silhouette score

```
range_n_clusters = [2, 3, 4, 5, 6]

for n_clusters in range_n_clusters:

    clusterer = KMeans(n_clusters=n_clusters, random_state=0).fit(table)

    silhouette_avg = silhouette_score(table, clusterer.labels_)

    print(
        "For n_clusters =",
        n_clusters,
        "The average silhouette_score is :",
        silhouette_avg,
    )
```

```
For n_clusters = 2 The average silhouette_score is : 0.9109035883498503
For n_clusters = 3 The average silhouette_score is : 0.789415217948282
For n_clusters = 4 The average silhouette_score is : 0.7210837816951361
For n_clusters = 5 The average silhouette_score is : 0.7092426089710087
For n_clusters = 6 The average silhouette_score is : 0.6346130603227677
```

- As we can see, we will select 2 as the number of clusters
- The model cluster the harassment 2 into the smallest cluster for all the races

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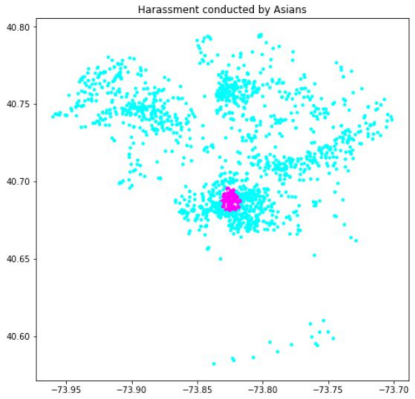
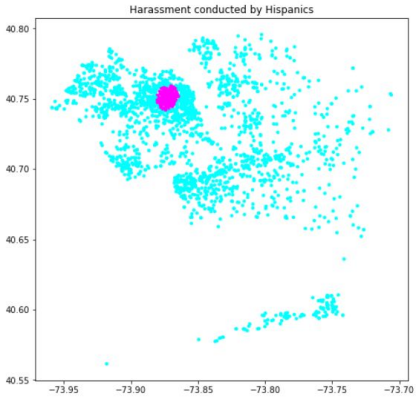
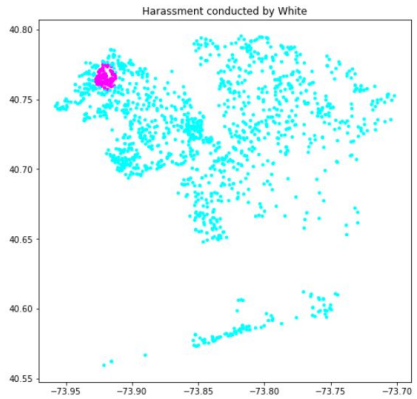
DBScan Analysis



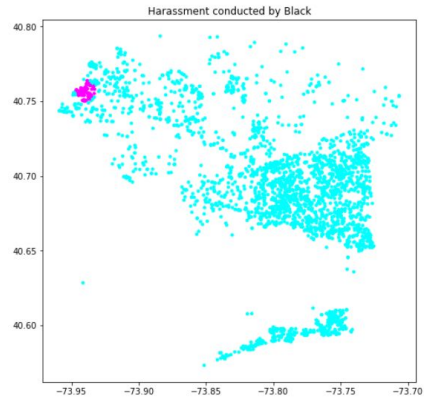
Asian



Hispanic



White



Black

- The lower part of Queens has the most serious asian harassment problem.
- Each race's most serious harassment problem happened in different places in Queens.

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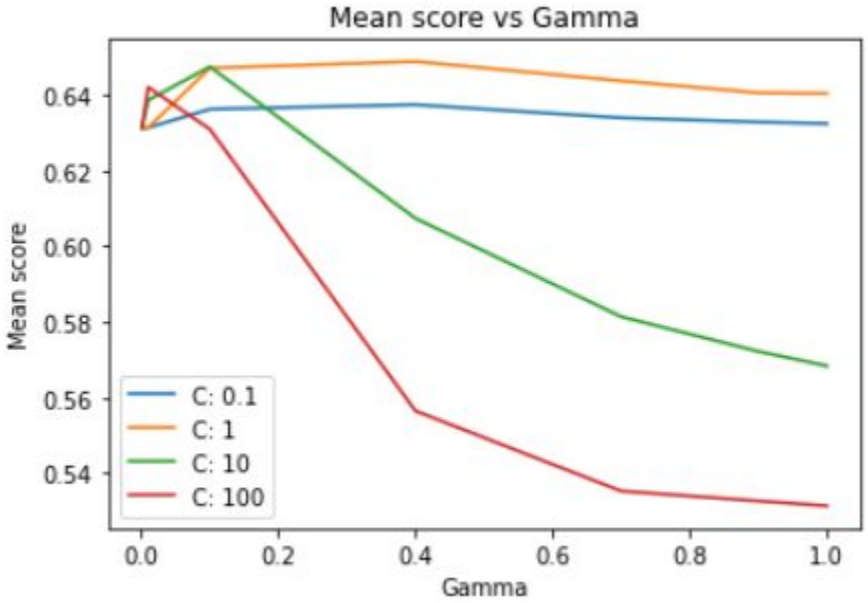
SVM




```
grid = GridSearchCV(svm.SVC(), param_grid=param_grid, cv=2)

grid.fit(X_train, Y_train)

print("The best classifier is: ", grid.best_estimator_)
```

```
The best classifier is:  SVC(C=1, gamma=0.4)
```

```
# plot the scores of the grid
# grid_scores_ contains parameter settings and scores
ypred1 = grid.predict(X_test)
print("Out of sample, rbf svm successfully predicts {} percent of the data".format(accuracy_score(Y_test,ypred1)
```

```
Out of sample, rbf svm successfully predicts 0.6562842528315674 percent of the data
```

```
grid.cv_results_['mean_test_score']
```

```
array([0.63242902, 0.63283395, 0.63395874, 0.63742294, 0.63625324,
       0.63134926, 0.63134926, 0.64039226, 0.64057224, 0.64372151,
       0.64885036, 0.64709588, 0.63134926, 0.63134926, 0.56836296,
       0.57218715, 0.58136519, 0.60736948, 0.64745568, 0.63863774,
       0.63134926, 0.53133583, 0.53264064, 0.53524996, 0.55644048,
       0.63094431, 0.64205699, 0.63130427])
```

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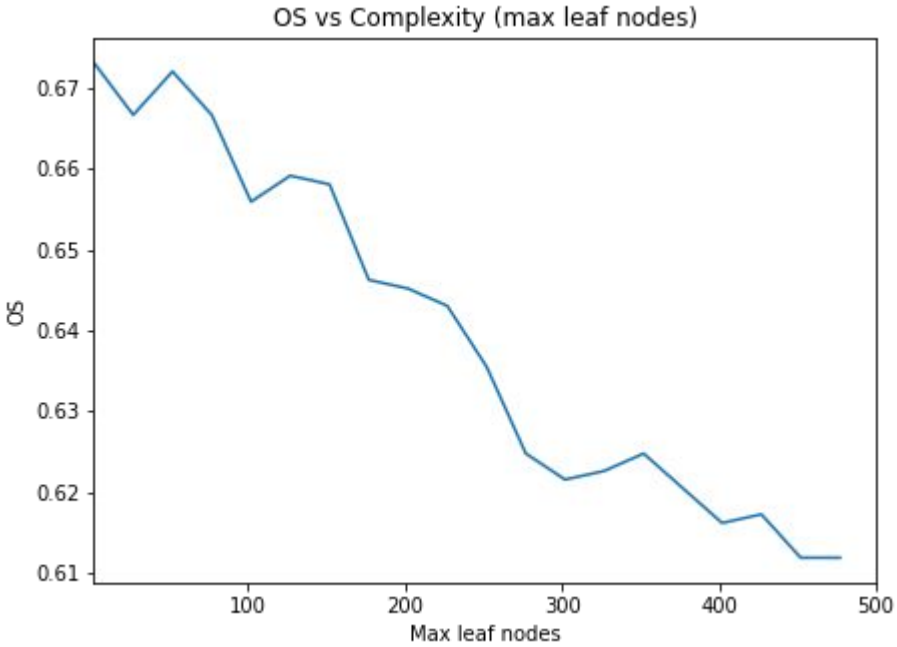
Decision Tree




```python
# your code here
from sklearn.model_selection import GridSearchCV

param_grid = {'max_leaf_nodes':range(2,500,25)}
dt=DecisionTreeClassifier(random_state=42)
gr=GridSearchCV(dt,param_grid=param_grid,scoring='accuracy')
rs=gr.fit(X_train,y_train)
print(rs.best_params_)
print(rs.score(X_test,y_test))
```

```
/usr/local/lib/python3.7/dist-packages/sklearn/model_selection
  UserWarning,
{'max_leaf_nodes': 27}
0.6666666666666666
```

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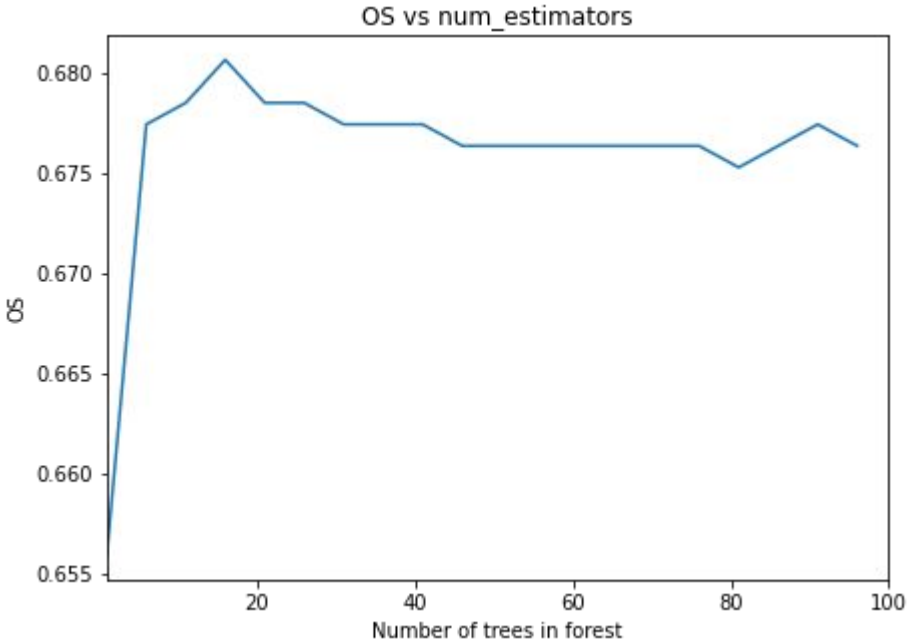
Random Forest




```
param_grid = {'n_estimators':range(1,100,5)}
rf = RandomForestClassifier(n_jobs=-1,max_leaf_nodes=27, random_state = 42)
gs = GridSearchCV(rf,param_grid=param_grid,scoring='accuracy')
rs = gs.fit(X_train,y_train)
pred=rs.predict_proba(X_test)[:,1]
print(rs.best_params_)
print(rs.score(X_test,y_test))
```

```
/usr/local/lib/python3.7/dist-packages/sklearn/model_selection/_split.py:680
  UserWarning,
{'n_estimators': 11}
0.678494623655914
```

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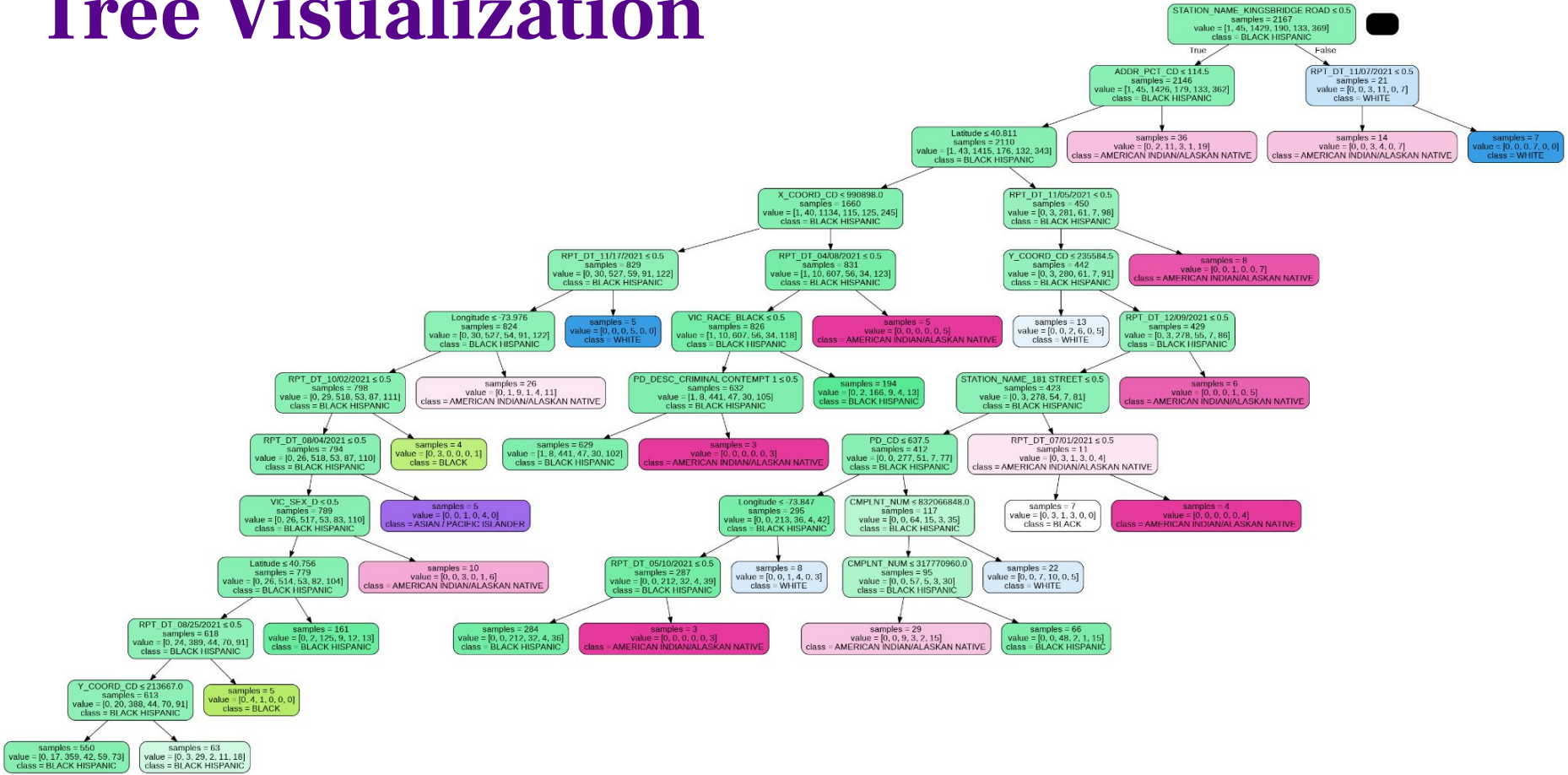
Tree Visualization



Anomaly Detection

Result of Gaussian Mixture:

OFNS_DESC	score
PROSTITUTION & RELATED OFFENSES	-6.653364
UNLAWFUL POSS. WEAP. ON SCHOOL	-4.762023
CHILD ABANDONMENT/NON SUPPORT	-4.579684
INTOXICATED/IMPAIRED DRIVING	-3.662789
FELONY SEX CRIMES	-2.970356

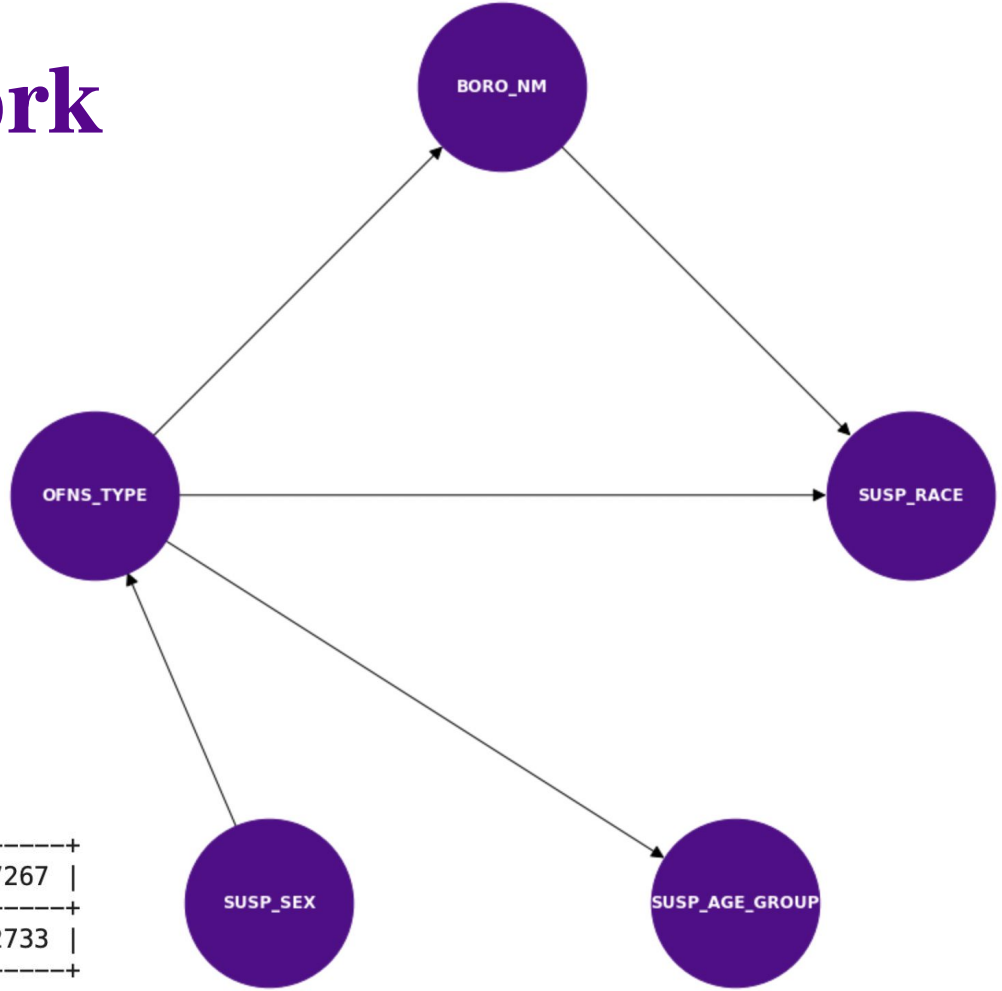
Result of K-Means:

OFNS_DESC	distance
PROSTITUTION & RELATED OFFENSES	1.480624
INTOXICATED/IMPAIRED DRIVING	1.065731
UNLAWFUL POSS. WEAP. ON SCHOOL	1.054180
CHILD ABANDONMENT/NON SUPPORT	1.013794
HOMICIDE-NEGLIGENT-VEHICLE	0.890717

PART 02

Bayesian Network

SUSP_SEX	SUSP_SEX(F)	SUSP_SEX(M)
OFNS_DESC(FELONY ASSAULT)	0.21	0.22
OFNS_DESC(HARRASSMENT 2)	0.71	0.6
OFNS_DESC(MISCELLANEOUS PENAL LAW)	0.08	0.17



CPD of SUSP_SEX:

SUSP_SEX(F)	0.27267
SUSP_SEX(M)	0.72733

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Conclusions

- Victims: Asians, American Indians are similar, Hispanics and Blacks are similar when they are victims
- SVM and Random Forest: decent accuracy could be improved by tuning more precisely
- Servere harassment is most popular for all races. Crime locations are discrete for all races
- Offense types affect community’s formation