

IOD Capstone Project: East London Police Use of Force Dataset

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About me and Agenda

- Joel Domingo
- 24 year year old Graduate
- Studied IT and Accounting at University of Technology Sydney (2020)



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Agenda:

- Introduction
- About the dataset
- Preliminary Insights
- Modelling
- Conclusions

Introduction

"Studies show law enforcement with some college education use force much less often than those with no higher education..."

"... only 1% of police forces within the United States have education requirements for those looking to join their forces."

Rydberg, Jason; Terrill, William (2010). "The Effect of Higher Education on Police Behavior". *Police Quarterly*. **13** (1): 92–120.

Main Stakeholders:

- Metropolitan Police Department (London)
- Police Agencies (worldwide)
- Civilians (their safety)
- Officers
- Government (funding changes)

Goal:

- To explore the **factors** which cause Police to use force, and **aim to reduce** that **internally** through resource allocation, and training and modelling.
- Create a model which can **predict whether use of force** may be **required**, and give **insights and prepare** officers responding to incidents.



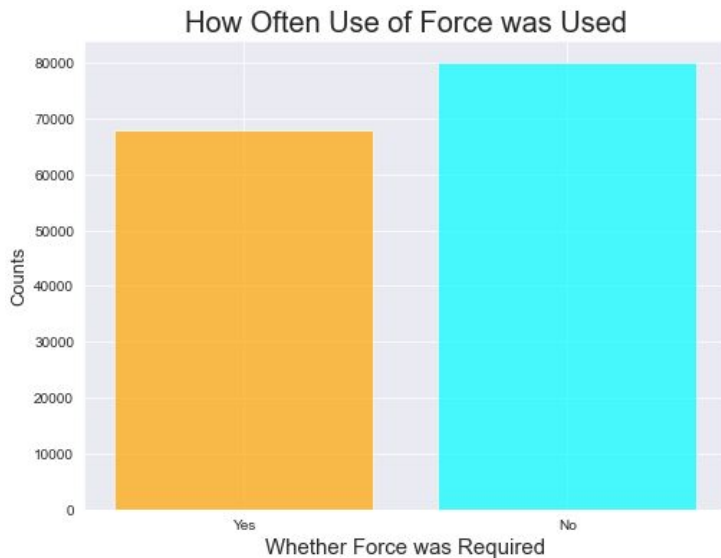


About the Dataset

- Dataset is a collection of features gathered from incident reports from the London Metropolitan Police Service, hosted by the London Government (source: <https://data.london.gov.uk/dataset/use-of-force>)
- Contains data ranging from 1st April 2020 to 31 January 2021.
- Contains 147,895 records of officer accounts, each with 271 features.



Insights: Use of Force Frequency



Q: What counts as 'Use of Force'?

- Anything beyond **compliant handcuffing** and **tactical communications**.

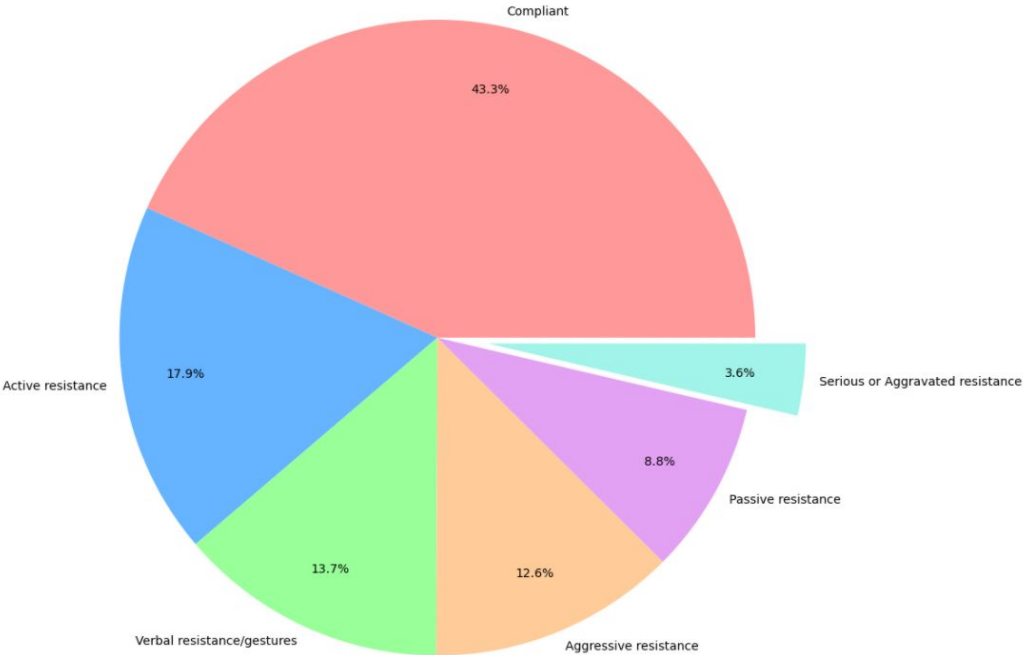
	Totals
Required Force	67,950
Did not require Force	79,945
Total	147,895



Insights: Subjects Analysis

Primary Conduct of Subject

Conduct	Counts
Compliant	64,058
Active Resistance	26,525
Verbal Resistance	20,243
Aggressive Resistance	18,655
Passive Resistance	13,017
Serious of Aggravated Resistance	5397

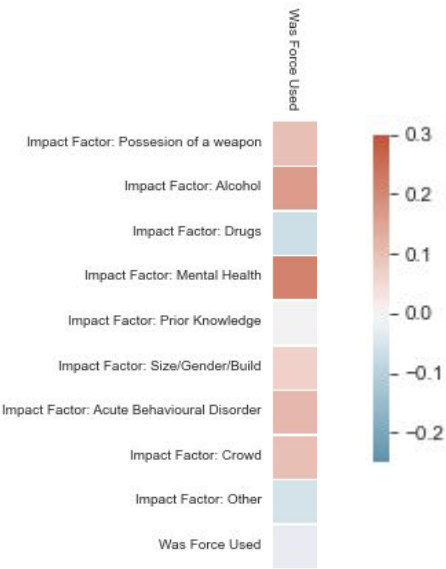


Further subject demographic analysis can be found in the provided report.

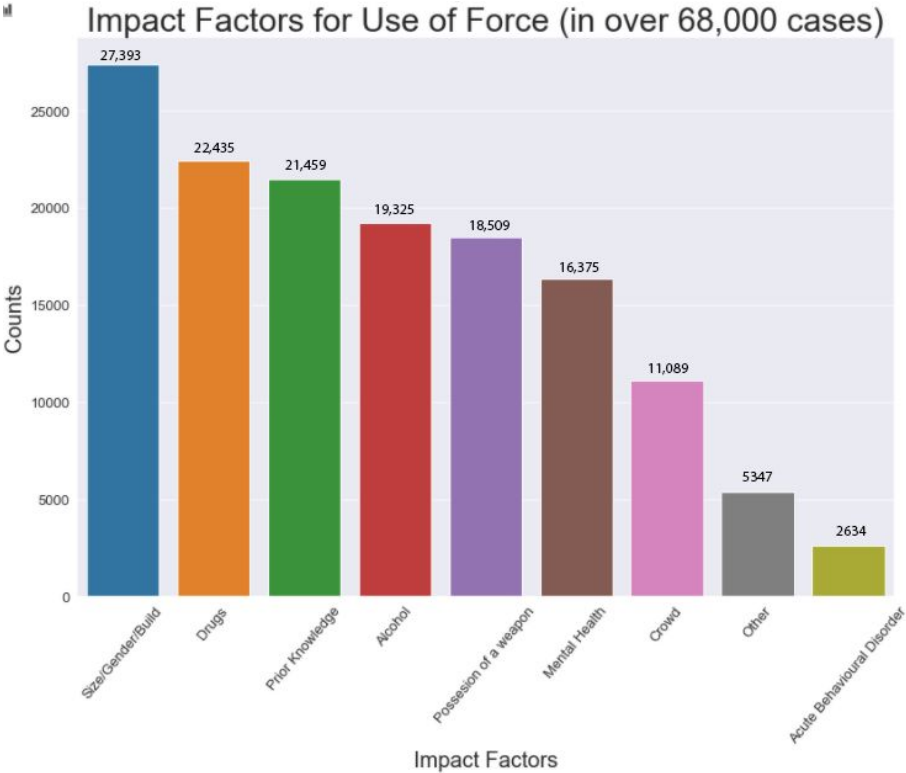


Impact Factors for Use of Force

Correlation:

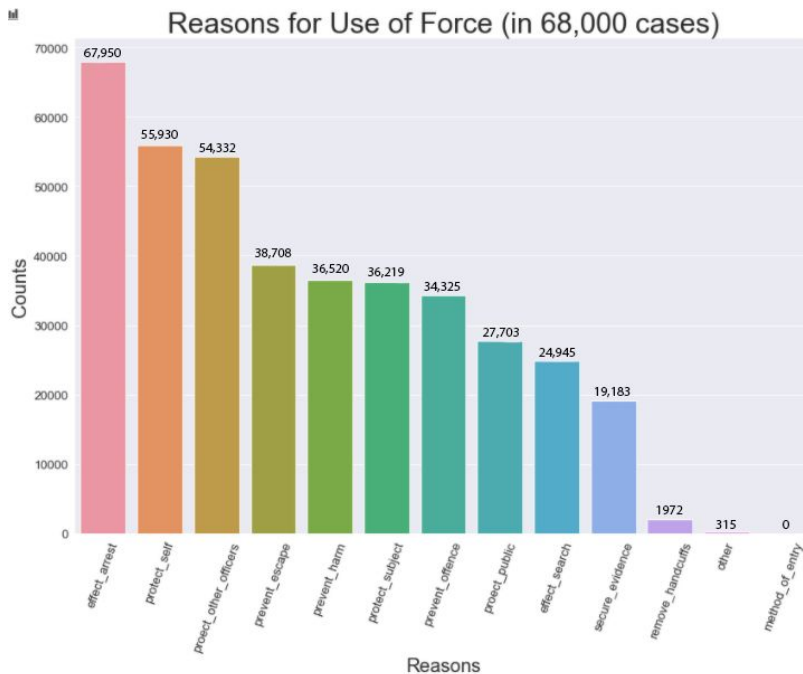


Impact Factors	Counts
Size/Gender/Build	27,393
Drugs	22,435
Prior Knowledge	21,459
Alcohol	19,235
Possession of Weapon	18,509

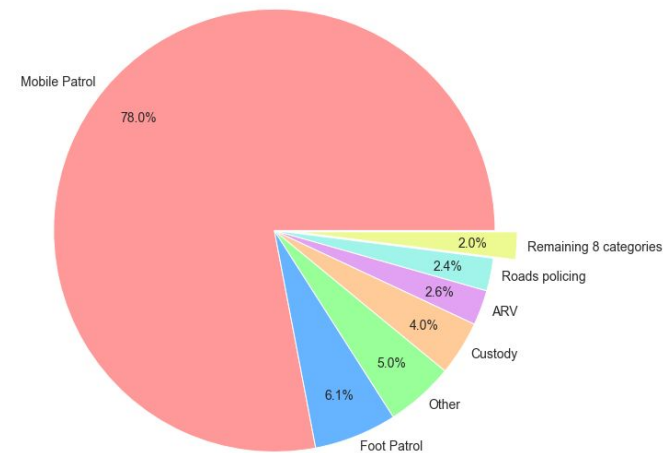


Why are police required to use force?

Reasons	Counts
Effect arrest	67,950
Protect self	55,930
Protect Other officers	54,332
Prevent Escape	38,708
Prevent harm	36,520
Protect subject	36,219
Prevent offence	34,325
Protect public	27,703
Effect search	24,945
Secure evidence	19,183
Other	315
Method of entry	0



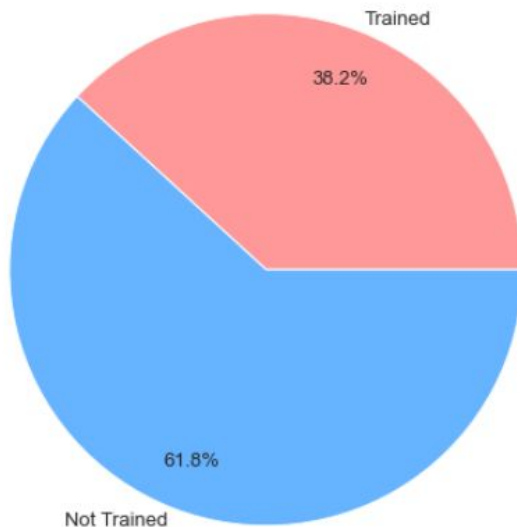
Proportion of Incidents Requiring Force per Main Duty





Taser Use

Proportion of Officers trained in CED (in 149,895 cases)



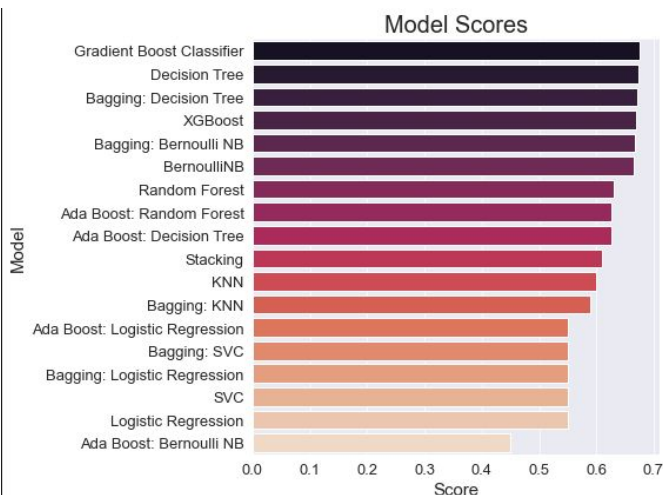
Trained in Taser?	Counts
Trained	56,450
Not Trained	91,445

Number of Officers trained in CED (Taser)

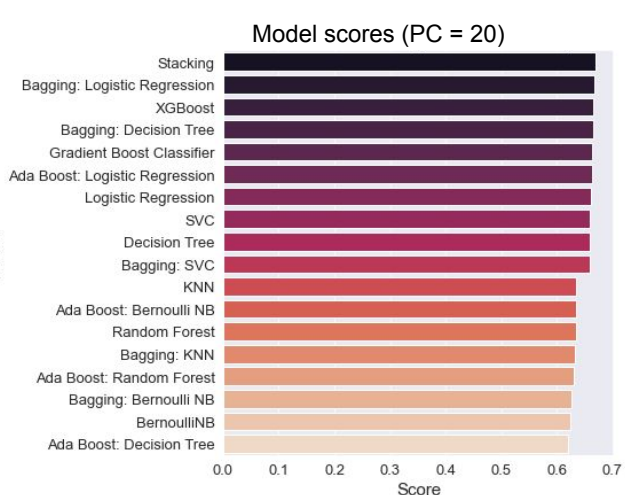


Modeling:

Scores	
Gradient Boost Classifier	0.676
Decision Tree	0.674
Bagging: Decision Tree	0.672
XGBoost	0.670
Bagging: Bernoulli NB	0.668
BernoulliNB	0.666
Random Forest	0.630
Ada Boost: Random Forest	0.627
Ada Boost: Decision Tree	0.626
Stacking	0.609
KNN	0.600
Bagging: KNN	0.590
Ada Boost: Logistic Regression	0.550
Bagging: SVC	0.550
Bagging: Logistic Regression	0.550
SVC	0.550
Logistic Regression	0.550
Ada Boost: Bernoulli NB	0.450



Scores	
Stacking	0.669
Bagging: Logistic Regression	0.667
XGBoost	0.666
Bagging: Decision Tree	0.666
Gradient Boost Classifier	0.663
Ada Boost: Logistic Regression	0.662
Logistic Regression	0.660
SVC	0.659
Decision Tree	0.658
Bagging: SVC	0.658
KNN	0.635
Ada Boost: Bernoulli NB	0.635
Random Forest	0.634
Bagging: KNN	0.632
Ada Boost: Random Forest	0.630
Bagging: Bernoulli NB	0.626
BernoulliNB	0.624
Ada Boost: Decision Tree	0.620



- Conducting dimensionality reduction reduced processing time slightly.
- It improved the average result of all models, with more homogeneous results.
- However, the top models were unable to exceed the results without dimensionality reduction.

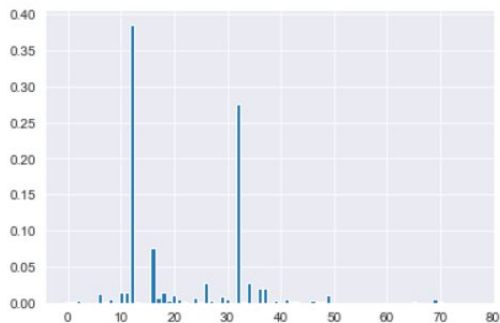


Final Model and Use Case - Decision Tree or Gradient Booster

Scores	
Decision Tree	0.662
Gradient Booster	0.662

Hypothetical Scenario: Officer responding to call with the following features. Using the models, predict whether the officer should expect/be prepared to use force on the subject.

- At Sports Stadium
- Officer called is mobile patrol
- They are Single Crewed
- subject age 24
- No disability
- Male
- Asian
- In Lewisham



	Outcome	Chance that Officer required to use force
Decision Tree	0	0.28
Gradient Booster	0	0.40



Conclusions and Some recommendations:

- Analysis revealed observations significant enough to make recommendations to police agencies
- Large frequency of use of force is largely factored towards **unpreparedness**
- The goal is to find ways to reduce risk to both officers and public regarding safety, and we can do this by increasing preparedness
- This model and analysis varies on an international scale, but the fundamental principles and features can be applied anywhere in the world.

Key Recommendations:

- Implement further training in tactical communications
- Implement education requirements prior to successful recruitment
- Implement further training on dealing with mentally-ill individuals
- Implement suggested model and distribute to all London Police departments.