Big Data Analytics: London Crime Data Analysis

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Overview

1 Introduction

- 2 Data Understanding
- 3 Cluster Analysis

The analysis' purpose

To discover the patterns among the criminal activities in the London metropolitan area in a distinct window of time.

The Dataset(1)

London Crime Data, 2008-2016: this dataset, hosted by **Kaggle**, is composed by 13 millions rows describing the London metropolitan area's criminal activities by *Borough*, *Category, Month* and *Year* in a window of time that ranges from January 2008 to December 2016.

The Dataset(2)

The dataset is composed by 7 variables:

- Isoa_code: code for Lower Super Output Area in Greater London;
- borough: common name for London borough;
- major_category: high level categorization of crime;
- minor_category: low level categorization of crime within major category;
- **year**: year of reported counts, 2008 − 2016;
- **month**: month of reported counts, 1 12;
- value: monthly reported count of categorical crime in given borough;

The Dataset(3)

The variables *Isoa_code*, *borough*, *major_category*, *minor_category*, *year* and *month* are **categorical** variables, while *value* is a **discrete numerical** variable.

Numeric Variables' Analysis(1)

value is the only numeric variables in the dataset, it represents the monthly reported count of categorical crime in given borough and has 247 unique values. Its minimum value is 0 and its maximum value is 309, the mode is 0, which appears in the 74.56% of the dataset's samples.

Numeric Variables' Analysis(2)

Since 10,071,505, that is, the 74.56% of the dataset's samples have the variable value eguals to 0, we can conclude that, on a superficial level, the window of time from 2008 to 2016 wasn't too dense of criminal activities.

Crimes per Year

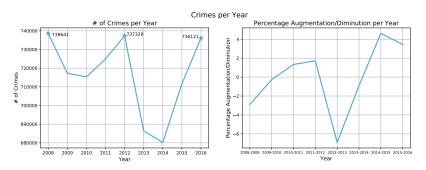


Figure: Crime's progress over the years

Crimes per Month

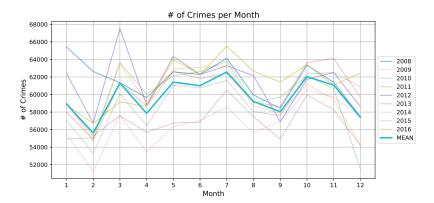
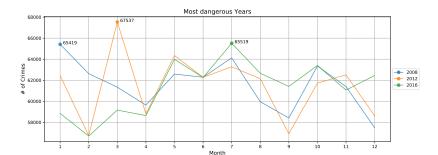


Figure: Crime's progress over the months

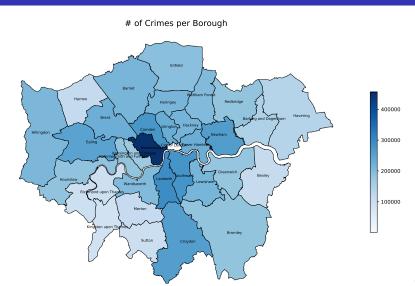
Most Dangerous Years



Categorical Variables' Analysis

- borough has 33 unique values, of which Lambeth is the most frequent, appearing in the 4.47% of the cropped dataset's records;
- major_category has 9 unique values, of which Theft and Handling is the most frequent, appearing in the 33.25% of the cropped dataset's records;
- year has 9 unique values, of which 2016 is the most frequent, appearing in the 11.45% of the cropped dataset's records;
- month has 12 unique values, of which 7 is the most frequent, appearing in the 8.66% of the cropped dataset's records;

Crimes per Borough

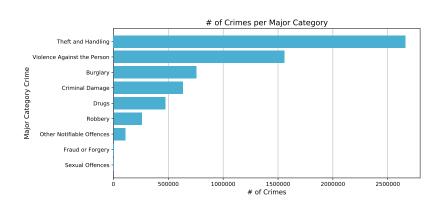


Crimes per Borough over Population Density

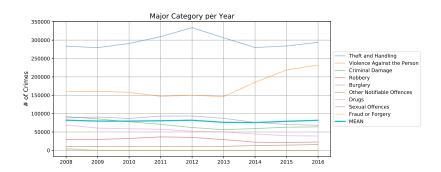
of Crimes per Borough over Population



Crimes per Major Category



Major Category Crimes per Year



Correlation Analysis

\$	Isoa_code \$	borough \$	major_category \$	minor_category \$	value \$	year \$	month \$
Isoa_code	D	D	D	D	D	D	- 1
borough	D	D	D	D	D	D	D
major_category	D	D	D	D	D	D	D
minor_category	D	D	D	D	D	D	D
value	D	D	D	D	D	D	D
year	D	D	D	D	D	D	D
month	1	D	D	D	D	D	D

Introduction

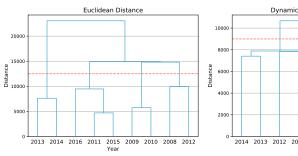
I have decided to enrich the informations provided with the data understaning by searching for possible **cluster-like structures** in the **time series** extracted from the main dataset, that is, hoping to discover the similarities (or dissimilarities) among the series describing the criminal activities from 2008 to 2016

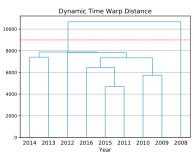
Choice of the Algorithms

I have used three popular clustering algorithms, that is, the KMeans algorithm, the Hierarchical Agglomerative algorithm and the DBSCAN. The three algorithms were adapted depending on the different series they were applied on.

By-Year Series: Hierarchical Agglomerative Clustering

By-Year Series: Hierarchical Agglomerative Clustering





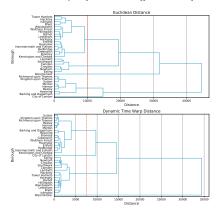
By-Year Series: KMeans Algorithm and DBSCAN Clustering

\$	Cluster \$	+	Cluster \$
Year \$	\$	Year ≑	\$
2009	0	2008	-1
2010	0	2012	-1
2011	1	2009	0
2015	1	2010	0
2016	1	2011	1
2008	2	2015	1
2012	2	2016	1
2013	3	2013	2
2014	3	2014	2

(a) KMeans (b) DBSCAN

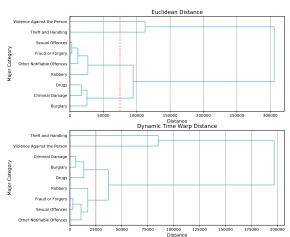
By-Borough Series: Hierarchical Agglomerative Clustering

By-Borough Series: Hierarchical Applomerative Clustering



By-Major Category Series: Hierarchical Agglomerative Clustering

By-Major Category Series: Hierarchical Agglomerative Clustering



By-Major Category Series: KMeans Algorithm and DBSCAN Clustering

+	Cluster \$	\$	Cluster \$
Major Category Crime \$	\$	Major Category \$	
Robbery	0	Theft and Handling	-1
Fraud or Forgery	0	Violence Against the Person	-1
Sexual Offences	0	Criminal Damage	0
Other Notifiable Offences	0	Burglary	0
Theft and Handling	1	Drugs	0
Violence Against the Person	2	Robbery	1
Criminal Damage	3	Fraud or Forgery	1
Burglary	3	Sexual Offences	1
Drugs	3	Other Notifiable Offences	1

(a) KMeans

(b) DBSCAN