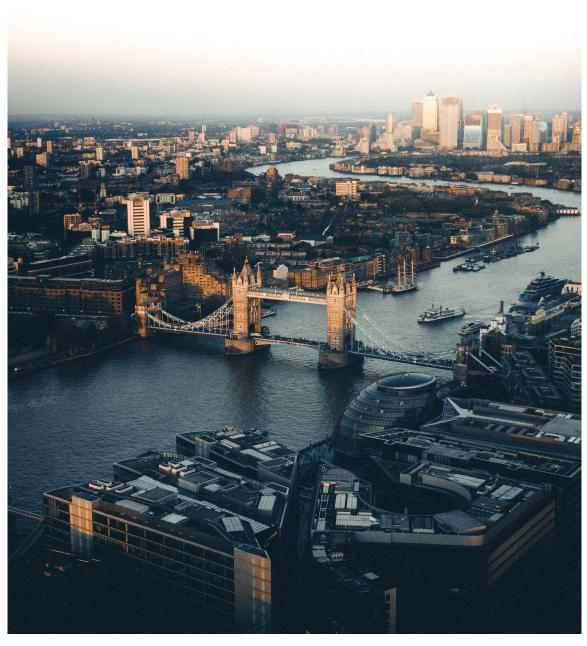
# Investigation into the relationship between local businesses and venues and crime in London

IBM Data Science Professional Certificate Capstone Project



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# **Table of Contents**

1.	Disclaimer	3
2.	Executive Summary	3
	Introduction	
	3.1 Problem Statement	4
	3.2 Scope	4
	3.3 Stakeholders	4
4.	Background	5
	Data	
	5.1 London Crime Data	6
	5.2 Local businesses/venues data	7
	5.2.1 Foursquare	7
	5.2.2 Venues with 24 Hour Licence	
	5.2.3 Local Units by industry	9
	5.3 Ancillary Data	
	5.3.1 London Population Data	10
	5.3.2 Geocoding	
	5.3.3 London maps	
	5.3.4 List of London Neighbourhoods	11
	5.3.5 List of London borough locations	
6.	Outline approach	
	Data Exploration	
	7.1 London Population Data	14
	7.2 London Crime Data	
	7.2.1 Crime trends over most recent 24 months	14
	7.2.2 Crime trends across minor crime category	14
	7.2.3 Crime trends across geography	
	7.2.4 Impact of population size	
	7.2.5 Ward level breakdown	19
	7.2.6 Crime data exploration conclusions	21
	7.2.7 Data transformation:	21
	7.3 Industry/Venue Data	21
	7.3.1 Foursquare	
	7.3.1.1 Data Obtained by Borough	21
	7.3.1.1.1 Using Central Points	21
	7.3.1.1.2 Using Borough within Foursquare API	21
	7.3.1.2 Data Obtained by Neighbourhood	22
	7.3.1.2.1 Using Central Points	
	7.3.1.2.2 Using Neighbourhood within Foursquare API	22
	7.3.2 24 Hour Licenced Venues	23
	7.3.2.1 Data transformation:	25
	7.3.3 Local Units by Industry	25
	7.3.4 Data transformation:	27
8.	Results	28
9	Conclusions	28

# 1. Disclaimer

This is not a formal analysis. It has been commissioned as a capstone project for IBM Data Science Professional Certificate. No reliance should be placed on conclusions and analysis within. This report has been produced under a number of constraints, including time and requirements of the capstone project and should not be considered fully representative of the quality of work the author can produce under professional conditions.

# 2. Executive Summary

[placeholder]

# 3. Introduction

London is the capital city of the United Kingdom, with an estimated population of 9.4 million<sup>1</sup>. Modern London is a diverse city, with varied neighbourhoods. Crime is a concern to residents, workers and business owners alike, in May 2021 a YouGov survey revealed that crime and housing were the biggest concerns to Londoners<sup>2</sup>.

The costs of crime can be direct ie for the victim of theft, or indirect, ie loss of business revenue, increased insurance. In the 2018 report *The economic and social impact of crime 2<sup>nd</sup> Edition*<sup>3</sup> the Home office included estimated costs of UK crime in 2015/16 to be £50 billion against individuals and £9 billion against businesses.

### 3.1 Problem Statement

This report will address the following question:

 Within the context of London, is there evidence of a relationship between the types of facilities/industries in a locality and crime?

# 3.2 Scope

#### In scope:

- Identification of overall trends
- Consideration of different categories of crime
- · Deep dive into the impact of 24 hour alcohol licensed premises
- Recommendations for follow-up analysis

#### Out of scope:

 Understanding the nature of a relationship ie if increased presence of certain businesses is in response to crime rate or a cause of it

#### 3.3 Stakeholders

Then intended stakeholders for this report are:

Stakeholder	Interest	Example
Local planners	for consideration in approving planning applications, and deciding planning strategies	If 24 hour licensed premises are shown to be associated with increased crime licenses applications for such licenses may be rejected
Local residents and business owners	For understanding potential impact of local facilities and businesses on crime rate	if certain venues are shown to be associated with increased crime they may wish to petition against such venues opening.
Prospective residents and business owners	For understanding potential crime profile on an area based on the local facilities and businesses	if certain venues are shown to be associated with increased crime, they may wish to avoid properties near to such such premises

 $<sup>1 \</sup>qquad https://worldpopulationreview.com/world-cities/london-population \\$ 

<sup>2</sup> https://yougov.co.uk/topics/politics/articles-reports/2021/05/06/housing-and-crime-most-pressing-issues-facing-new-

<sup>3</sup> https://assets.publishing.service.gov.uk/government/uploads/system/uploads/attachment\_data/file/954485/the-economic-and-social-costs-of-crime-horr99.pdf

# 4. Background

#### **Geographic Considerations**

Insight into London's history provides context for the formation of neighbourhoods and political boundaries against which data is collected. London is a diverse and a historic city, first referenced as Londinium in the 2<sup>nd</sup> Century AD. It is the largest city in the UK, and currently, by population the 32<sup>nd</sup> largest in the world (source: worldometer). London was subject to rapid growth associated with industrialisation the 19<sup>th</sup> Century AD, when the population grow fivefold, and immigration in the late 20<sup>th</sup> century following mid century decline. On the whole, the growth of London has been organic rather than planned. Many neighbourhoods feature 19<sup>th</sup> century street layouts, interspersed with 1940s bomb damage and 1960s slum clearance and redevelopment. Redevelopment in old and new neighbourhoods continues to this day.

In addition to population growth the geographical area considered London has grown to incorporate what were once separate settlement boundaries. Notably in 1965, London was redefined, and what were parts of neighbouring counties became included as London boroughs This history has resulted in a diverse geography, with neighbourhoods not necessarily fitting within the political boundaries by which London is governed. Many London neighbourhoods represent historic parishes, once separate to London., others have developed around transport links, for example the London underground stations.

Today London is governed by division into 32 local government districts referred to boroughs each governed by a London borough council..These make up the ceremonial county of Greater London. The "City of London", the historic centre of London, is a separate ceremonial county. However, the two counties together comprise the region of Greater London, all of which is also governed by the Greater London Authority. This is the definition of London used in this report.

Each London borough is subdivided into wards. These wards represent areas of population and seats in the borough council, they are not distinct communities and can be residential with commercial centres and amenities in nearby wards.

#### Crime data Collection

Greater London is covered by three separate police forces, which collate crime data

Police Force: Responsible for policing of: The vast majority of London

City of London Police City of London

British Transport Police The national rail network and the London Underground

#### **Crime Categorisation**

Different stakeholders will primarily be concerned with different categories of crime. The economic and social impact of crime 2<sup>nd</sup> Edition captured particular concern for businesses: Thefts from businesses make up almost 90% of business crime but account for approximately half of the total estimated costs of crime against businesses (£4.2bn), as each crime has a low impact on society. In contrast, robberies and burglaries against businesses – estimated to cost £2bn and £1.6bn respectively – make up over 40% of the costs of crime, but account for only 5% of all crimes against businesses.

## 5. Data

Two primary sources of data will be used for this report:

- Foursquare: an independent location data platform, which harvest details and recommendation of venues as input by users of it's mobile application "Foursquare City Guide"
- 2. **The London Datastore:** a free and open data-sharing portal, provided by the London Assembly and Mayor of London.

This reports utilises three groupings of data:

- a. Crime
- b. Local businesses/venues
- c. Ancillary data to support analysis

#### 5.1 London Crime Data

**Data source:** The London Datastore. *Recorded Crime Summary: Geographic Breakdown*. Provided by the Metropolitan Police

https://data.london.gov.uk/dataset/recorded\_crime\_summary

#### **Summary of data:**

MPS Borough level crime (most recent 24 months): Number of offences by month, by category (both major and minor), by borough for 24 months (the most recent that data

is available for)

MPS Ward level crime (most recent 24 months): Number of offences by month, by category (both major and minor), by ward for 24 months (the most recent that data is

available for)

Borough level crime Number of offences by month, by category (both major and

(historic): minor), by borough for 2008 to 2018

	MajorText	MinorText	LookUp_BoroughName	201905	201906	201907	201908	201909	201910	201911		202007	202008	202009	202010	202011	202012	202101	202102	202103	202104
0	Arson and Criminal Damage	Arson	Barking and Dagenham	11	3	5	3	6	9	8		4	6	2	7	4	2	4	6	4	6
1	Arson and Criminal Damage	Criminal Damage	Barking and Dagenham	140	113	134	118	109	109	97		122	114	116	120	100	109	100	104	80	100
2	Burglary	Burglary - Business and Community	Barking and Dagenham	21	27	31	35	37	30	30		28	23	32	21	18	24	20	18	14	12
3	Burglary	Burglary - Residential	Barking and Dagenham	114	96	71	67	80	97	114	***	72	63	54	68	90	91	69	90	71	75
4	Drug Offences	Drug Trafficking	Barking and Dagenham	9	6	-11	8	7	9	14		21	9	12	13	17	13	12	9	7	6

Figure 1: MPS Borough Level Crime (most recent 24 month). First 5 records

See appendix for details of crime categories

Data cleansing: None required, data is complete and good quality.

**Data limitations:** Data used is crime statistics published by the Metropolitan police so will exclude the City of London, and crime recorded on the London underground and national rail network.

# 5.2 Local businesses/venues data

This report will consider three different sources of data reflecting the facilities and venues. The London datastore contains additional relevant data which could be used for follow-up research.

# 5.2.1 Foursquare

#### Data source:

GET https://api.foursquare.com/v2/venues/explore

Foursquare API "explore" endpoint returns up to 50 recommended venues. These venues are those from a geographic area either specified by a

- geocodable location name, or
- longitude/latitude coordinates provided with a search radius.

The underlying Foursquare places database is crowd sourced.

The most meaningful approach to harvesting Foursquare venue data for this analysis will be finalised post data exploration. Options include by name or coordinates and by neighbourhood or borough

#### Summary of data:

Foursquare API returns a JSON response, details of which can be found on the developer reference guide.

For this exercise the

- venue name
- venue location (latitude, longitude)
- venue category

will be considered.

The category provides a description of the type of venue and allows grouping of similar venues for further analysis

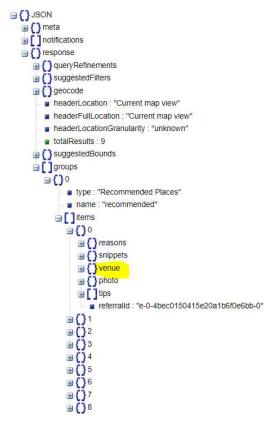


Figure 2: Example Foursquare response

<sup>4</sup> https://developer.foursquare.com/docs/api-reference/venues/exp



Figure 3: Example Foursquare response - venue node expanded

Data cleansing: Not required

Data limitations: There are a number of limitations with foursquare data

- 1. Crowd sourced data is biased by the demographic which uses foursquare
- 2. The geographic area data is obtained for may not match, or be representative of the geographic area crime data is available for
- 3. The limit of responses may prevent the returned venues being representative of the area searched. This limit is 50 as per documentation, but experience suggests is actually 100.
- 4. The allocation of category can be subjective, for example coffee shop vs cafe, pub vs gastropub, fried chicken joint vs fast food restaurant.

#### 5.2.2 Venues with 24 Hour Licence

This data allows detailed analysis into a particular venue type.

**Data source:** The London Data Store *Alcohol and late-night refreshment licensing statistics – licensed premises 24 hour.* Provided by the Home Office. https://data.london.gov.uk/dataset/alcohol-and-late-night-refreshment-licensing-statistics

## Summary of data:

Data is provided on an annual basis, with the most recent data being from 2018.

Per year the number of licensed premises is provided for each borough (licensing authority), broken down by premises type with totals and subtotals across categories of premises.

Propose to use 2018 data with 2017 data used to fill in gaps

	Premises with 24-hour alcohol licences													
			Supermarkets and stores Hotel bars											
Licensing authority	Total	Pubs, bars and nightclubs	Total	Large supermarket	Other convenience stores	Supermarket and store type not reported	Total	Open 24 hours to residents and general public	Open 24 hours to residents and their guests only	Hotel bar type not reported	Other premises types	Premises type not reported		
Barking and Dagenham	:	:	3	1	2	0	:	0	:	3	0			
Barnet	:	:	3	:	:	:	:	:	:	3	:			
Bexley	4	0	4	1	3	0	0	0	0	0	0	(		
Brent	46	2	28	4	24	0	3	3	0	0	13	(		
Bromley	14	0	11	4	7	0	3	0	3	0	0	(		
Camden	:		:	:		:	:	:	:		:			

Figure 4: Example data on licensed premises (24 hour license)

#### Data cleansing:

Data is cleansed to enable analysis:

- There are a number of gaps in the data (shown as :), where that data is not available. Use the previous years' value if available.
- Ensure totals and subtotals reflect any data cleansing
- If any data is missing for 2 consecutive years exclude
- Exclude any features (columns) where more than eight records are 0, as there will be insufficient data for meaningful analysis



Figure 5: Example cleansed data showing feature reduction

# **5.2.3 Local Units by industry**

#### Data source:

The London datastore: *Local unity by broad industry group, borough. P*rovided by the Office of National Statistics.

https://data.london.gov.uk/dataset/local-units-broad-industry-group-borough

#### Summary of data:

Data is available for each year 2003 to 2020. The number of businesses (local units such as a factory or a shop) by Broad Industry Groups, per borough is provided per year.

UK SIC 2007																			
Code	Area	SIC07: 01-03: Agriculture , forestry & fishing	SIC07: 05-39 : Production	SIC07: 41-43 : Constructio n	SIC07: 45: Motor trades	SIC07: 46 : Wholesale	SIC07: 47 : Retail	SIC07: 49-53: Transport & Storage (inc. postal)	SIC07: 55-56 : Accommodatio n & food services	SIC07: 58-63 : Information & communicatio n	SIC07: 64-66: Finance & insurance	SIC07: 68: Property	SIC07: 69-75: Professional , scientific & technical		SIC07: 84 : Public administratio n & defence	SIC07: 85 : Educatio n	SIC07: 86-88: Health	SIC07: 90-99 : Arts, entertainment , recreation & other services	SIC07: Total
E09000001	City of London	20	730	725	25	605	995	285	1,280	2,185	3,700	1,055	8,005	6,105	55	215	330	1,050	27,365
E09000002	Barking and Dagenham	10	330	1,460	225	410	645	530	410	610	90	145	810	780	170	200	605	360	7,790
E09000003	Barnet	30	625	3,200	350	1,155	2,165	600	1,000	2,570	530	2,055	4,910	2,205	55	535	1,250	1,585	24,820
E09000004	Bexley	10	445	1,905	260	335	805	380	570	1,065	180	225	1,510	840	25	225	555	560	9,895
E09000005	Brent	5	585	2,255	410	955	1,600	670	915	1,870	270	755	2,630	1,355	60	320	790	1,015	16,460
E09000006	Bromley	60	495	2,355	290	525	1,665	310	855	2,080	390	510	3,465	1,570	25	380	900	1,195	17,070
E09000007	Camden	30	1,005	1,515	150	1,240	2,450	465	2,140	4,610	855	1,525	10,385	3,530	115	680	1,205	2,845	34,745
E09000008	Croydon	15	495	2,285	390	540	1,540	500	1,020	2,030	335	525	2,930	1,445	50	420	1,095	1,000	16,615

Figure 6: Example local units data as available from London Data Store

#### Data cleansing:

Data is cleansed to enable analysis:

- Additional UK regional data is included in the dataset, records removed as only London Borough data is required
- No gaps identified in data

#### **Data limitations:**

Number of units provides a single snapshot, in reality businesses are not static, opening and closing through the year.

# 5.3 Ancillary Data

The following data is utilised to support analysis

# **5.3.1 London Population Data**

Population data enables the primary data to be population adjusted – eg crime rate per 10,000 capita to be calculated.

#### Data source:

London Datastore: London Borough Profiles. Provided by the Greater London Authority

https://data.london.gov.uk/dataset/london-borough-profiles

London Datastore: London Ward Profiles: Provided by the Greater London Authority

https://data.london.gov.uk/dataset/ward-profiles-and-atlas

#### Summary of data:

Extensive indicators are provided about each London ward and borough. Of interest to this analysis is population.

The population figure provided is: 2017 Estimate: Borough level data 2015 Estimate: Ward level data

	Code	Area_name	Inner/_Outer_London	GLA_Population_Estimate_2017	GLA_Household_Estimate_2017	Inland_Area_(Hectares)	Population_density_(per_hectare)_2017	Average_Age,_2017	Proportion_of_population_agi 15,_
0	E09000001	City of London	Inner London	8800	5326	290	30.3	43.2	
1	E09000002	Barking and Dagenham	Outer London	209000	78188	3,611	57.9	32.9	
2	E09000003	Barnet	Outer London	389600	151423	8,675	44.9	37.3	
3	E09000004	Bexley	Outer London	244300	97736	6,058	40.3	39.0	
4	F09000005	Brent	Outer London	332100	121048	4 323	76.8	35.6	

Figure 7: Header of Borough profile data

#### **Data limitations:**

It is assumed that population growth has been consistent across boroughs/wards since the date of the population estimates, therefore the population number is a reasonable proxy to use for current population.

# 5.3.2 Geocoding

Nominatim API is used to obtained coordinates of locations, based on a place name (ie London boroughs and neighbourhoods.

https://wiki.openstreetmap.org/wiki/Nominatim

This is used for data visualisations, and is investigated for utilising the Foursquare API to obtain venue data for a specific geography.

## **5.3.3 London maps**

The python Folium library is used for visualisations <a href="http://python-visualization.github.io/folium/">http://python-visualization.github.io/folium/</a>

London boroughs are overlaid using <code>london\_boroughs\_proper.geojson</code> sourced from: <a href="https://joshuaboyd1.carto.com/tables/london\_boroughs\_proper/public/map">https://joshuaboyd1.carto.com/tables/london\_boroughs\_proper/public/map</a>
London wards are overlaid using <code>london-wards-2014.geojson</code> sourced from: <a href="https://github.com/ft-interactive/geo-data">https://github.com/ft-interactive/geo-data</a>

# 5.3.4 List of London Neighbourhoods

Wikipedia provide a list of London neighbourhoods is scrapped from 'https://en.wikipedia.org/wiki/List\_of\_areas\_of\_London'



Figure 8: List of London neighbourhoods as shown on wikipedia

# 5.3.5 List of London borough locations

The wikipedia list of London boroughs provides coordinates for each <a href="https://en.wikipedia.org/wiki/List\_of\_London\_boroughs">https://en.wikipedia.org/wiki/List\_of\_London\_boroughs</a>



Figure 9: List of London boroughs (in part) as shown on wikipedia

# 6. Outline approach

Exploratory data analysis will confirm the data transformations required which may include:

- data aggregation (ie crime data over a time period)
- adjusting data for population (ie so crime/venues per 10,000 capita is considered)
- exclusion of outliers

Exploratory data analysis will also consider trends in different levels of crime data categorisation classification (total, major category, minor category) to determine at which level analysis should be conducted.

Foursquare data will be used to form clusters of geography (either neighbourhood or borough) based on similar venue profiles. Venue profiles will be determined using normalisation of number of venues of different categories in a geography. K-means clustering will be used to identify clusters of similar venue profiles. Map based visualisations will be used to enable manual inspection for correlations. Ie cluster labels superimposed on a choropleth map of crime data. Data exploration will confirm the geography used in this analysis.

Where crime and business data exists for the same geography, pairwise pearson correlation analysis will be conducted to identify where correlation exists.

This analysis will be conducted in python, libraries used will include:

BeautifulSoup for web scraping

numpy to handle data in a vectorized manner

pandas for data analysis

geopy.geocoders Nominatim to convert an address into latitude and longitude values

sklearn.cluster for k-means clustering folium for map rendering matplotlib for visualisations

seaborn for visualisations