

CONCEPT

- Ever since the UK government introduced new planning policy guidelines aimed at revitalising city centres and control out-of-town developments, planners and developers have seen cinemas as key amenities to make urban masterplans and their associated schemes more attractive.
- As major retailers come under pressure to attract shoppers with leisure-related activities, a growing number of cinema construction projects are being built across the country to help ensure that new developments succeed.
- The focus of this project will be establishing which Boroughs in the Greater London Area are underserved by both established cinema chains alongside smaller independent cinemas.

DATA SOURCES

- Office of National Statistics population data will be used to establish the population levels within each
 of the 32 London boroughs.
- Wikipedia entry for the London Boroughs will be utilised to gain an oversight of the names of both Boroughs and Local Authorities that control those Boroughs
- Utilising the Foursquare API to get an understanding of the number of cinemas already present within each Borough currently.

DATA SCRAPED FROM WIKIPEDIA

List of London boroughs From Wildpedia, the free encyclopedia This is a list of focal authority dauticits within Greater London, including 32 London boroughs and the encyclopedia This is a list of focal authority dauticits within Greater London, including 32 London boroughs and the removal part of London Boroughs and the removal of Cox and Plans Chy of Justice. For planning purposes, in addition to the boroughs and chy three are also two active development corporations, the London Longon Chy three are also two active development corporations, the London Longon Chy of List of Borough and local authorities [eat] **Extends finish** **List of boroughs and local authorities [eat]** **Barking and Dagentium [london Borough Council Labour Town Half. 1 Town Square 13:59 212:50 0 0 0 1507771 25 0 15077771 25 0 15077771 25 0 1507771 25 0 15077771 25 0 15077771 25 0

- To produce a list of each of the London Boroughs, in addition to further information about their political control, data was scraped using from a table on Wikipedia using Urllib.
- Once the data was scraped and assigned to a variable, we used Beautiful Soup to extract and work with the data in it.
- The class ID of the important tables on the page was 'Wikitable sortable' so a further function pulled out only tables with this ID
- Looped through the rows of the table using the tri (row) and td (cell) fields to populate the a Dataframe.

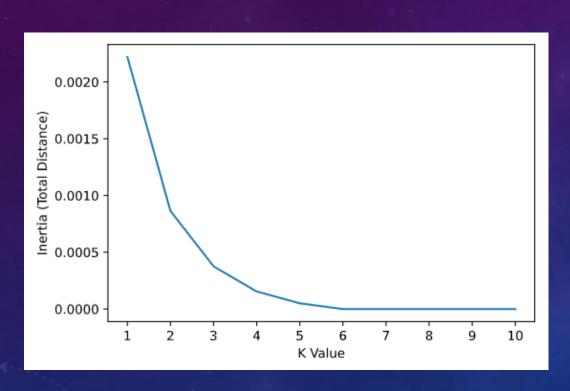
ADDING GEOGRAPHIC COORDINATES

- Once a table had been created and stored the data for each of the London Boroughs Geocoder was used to find the Latitude and Longitude of each of the geographic centres of the Boroughs.
- Source table did include Lat/Long information, this was for the political headquarters of each Borough, rather than the centre, the latter being most suitable for the analysis. This was saved as a separate Dataframe.
- Thus, Geocoder was used to find the Latitude and Longitude of each of the geographic centres of the Boroughs.

USING THE FOURSQUARE API

- These geographic coordinates were then passed through the Foursquare API to look for nearby venues within a 6 Kilometre range from the centre of each borough.
- A onehot was then produced to establish the mean number of each venue category produced by the resultant search and this Dataframe was then queried for unique values within 'Venue Category'
- This established 'Movie Theatre' and 'Indie Movie Theatre' as the categories of interest for this study.
 Onehots for these categories were extracted for each borough and saved as a new Dataframe.

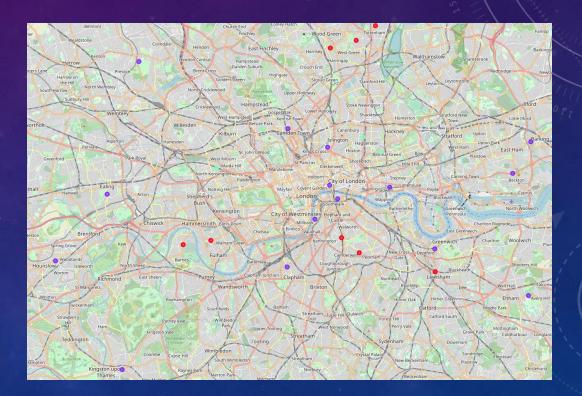
K-MEANS — CLUSTERING USING MACHINE LEARNING



- K-means Clustering was then utilised to cluster boroughs based on the output around the venue categories previously established.
- Prior to a full run, a Cluster Variance function was defined to establish inertia looking at total Euclidean distance between data values for K=1 through K=10.
- A K value of 3 was ultimately used as this produced three clear clusters, one with presence of either a Movie Theatre, an Independent Movie Theatre or neither.

K-MEANS CLUSTERS

- Three clusters divided the dataset into three distinct categories.
 - Boroughs well served by independent cinemas without any real coverage of mainstream chains and the influence brought by size of location, both in terms of number of screens and maximum footfall.
 - Boroughs with good coverage of mainstream cinemas, without a great deal of coverage of independent locations, and the ability for those locations to show the mixture of blockbuster and arthouse films that make them so well loved by their clientele.
 - Boroughs that are completely unserved by either type of location and might possibly present good investment opportunities by developers or chains looking to open new locations. This however leads on to the next analysis.



CONSIDERING POPULATION DATA

no theater = london_merged.loc[london_merged['Cluster Labels'] == 1] borough_joined = pd.merge(borough_table, no_theater, on='Borough', how='inner') borough_joined.drop(['Indie Movie Theater', 'Movie Theater'], axis=1, inplace=True) borough_joined.sort_values(['Population (2019 est)'], ascending=False, inplace=True) borough_joined											
_	Borough		Local a	uthority	Political control	Area (sq mi)	Population (2019 est)	Latitude_x	Longitude_x	Cluster Label	
10	Newham	Newham London	Borough	Council	Labour	13.98	353,134	51.5077	0.0469		
	Ealing	Ealing London	Borough	Council	Labour	21.44	341,806	51.5130	-0.3089		
14	Wandsworth	Wandsworth London	Borough	Council	Conservative	13.23	329,677	51.4567	-0.1910		
	Lambeth	Lambeth London	Borough	Council	Labour	10.36	326,034	51.4607	-0.1163		
11	Southwark	Southwark London	Borough	Council	Labour	11.14	318,830	51.5035	-0.0804		
	Greenwich	Greenwich London	Borough	Council	Labour	18.28	287,942	51.4892	0.0648		
13	Waltham Forest	Waltham Forest London			Labour	14.99	276,983	51.5908	-0.0134		
	Hounslow	Hounslow London			Labour	21.61	271,523	51.4746	-0.3680		
15	Westminster		ter City		Conservative	8.29	261,317	51.4973	-0.1372		
	Havering	Havering London			Conservative	43.35	259,552	51.5812	0.1837		
4	Harrow	Harrow London			Labour	19.49	251,160	51.5898	-0.3346		
	Bexley	Bexley London			Conservative	23.38	248,287	51.4549	0.1505		
		Islington London			Labour	5.74	242,467	51.5416	-0.1022		
	Barking and Dagenham Sutton	Barking and Dagenham Londor Sutton Londor			Labour Liberal Democrat	13.93 16.93	212,906 206,349	51.5607 51.3618	0.1557 -0.1945		
12											

- Arguably the most significant factor in deciding where to open a new cinema might be the potential number of customers that would help any newly opened location reach the critical mass of business required to transition into an established site.
- Overlaid ONS population data on cluster group completely unserved by either major chains or independent cinemas.
- This indicated that locations such as Newham or Ealing are best placed as areas of further exploration.

LIMITATIONS OF STUDY

Geographic coverage

- Searches cantered around a single geographical point for each of the boroughs with a set radius leaves open the possibility that a particular location, geographically proximate to two of the centres of each borough could well have been identified twice
- Conversely, it is entirely possible that a location was outside of the 6-kilometre radius that represented the search zone of venues around those Borough centroids.
- Worthwhile expanding the search radius, but then completing extra work to remove duplicates from the study. Perhaps assigning a location identified multiple times to the centroid that it is closest to, rather than either counting multiple times, or not at all.

Comparing potential locations

- Reasonable to suggest that any of the top 5 or 10 locations would make suitable candidates for a new cinema location.
- Therefore useful to conduct further analysis on additional complications such as average travel times to and from each location to establish the ease of access to any new site using existing public transport links.

IN SUMMARY

- It stands to reason that markets are poised to return to some semblance of normality in a post-Covid world.
- Businesses are certainly planning their exit strategy which may well involve expanding their footprint to include additional sites in underserved communities.
- It is with this fact in mind that this report focused on the integration of data science and machine learning technics to assist decision makers in making informed and efficient choices.
- This study was narrow in scope and certainly would require a significant expansion to provide any
 results that would be actionable in a real-world scenario.
- It could be reasonable to state that from the results provided, we could potentially recommend potential locations for future research into their theoretical profitability as venue locations.