

# Leveraging location and competitive datasets to evaluate business opportunities in Central London

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Professional Certification

# Background & Context

A young entrepreneur has been lucky enough to secure funding for their new business, a cafe!

They are based in London, where it is difficult to find an ideal location due to the saturation of the market.

However, they have enlisted support to take a data driven approach, to get the edge in this competitive landscape.

This following slides will outline the work conducted to find the location for their new cafe - aiming for a vibrant area within central London.

The analysis will factor in information about the competitive landscape and leverage various data science techniques to inform the final recommendations.

# Data

## Sources:

- Location Data - the final approach defined a grid of potential locations
- Market Data - the Foursquare API provided detailed information about existing venues

## Data Preparation

- The focus was on complementary and competitive venues, so the venue data was filtered and cleaned according to some rough definitions:
  - Complementary: bars, restaurants
  - Competitive: cafes
- To find a vibrant location in the city centre, the centrality of candidate locations was factored in, and whether they are more likely to be an entertainment destination or a quiet residential area.

# Methodology

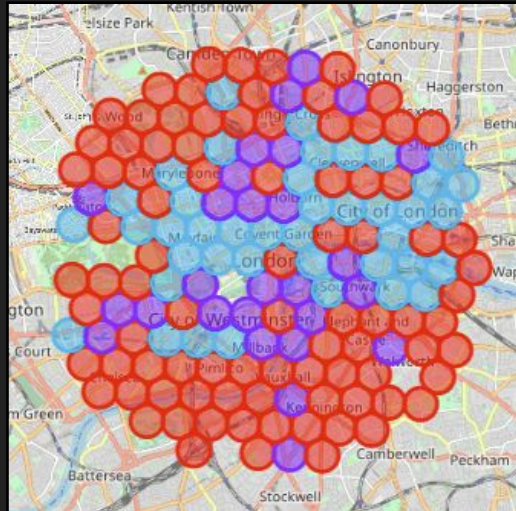
The investigation was approached in a number of steps:

1. A data frame was created with candidate locations
2. The Foursquare API was then used to provide information about the local venues, for each of the candidate locations, with this information a few analyses were conducted:
  - a. Calculated the number venues in the area
  - b. Cleaned the venue categories, to provide a manageable number of relevant categories for analysis
  - c. Understanding the distribution of locations along some key dimensions
  - d. Classifying locations, using a k-means algorithm (including the decision as to the best value of k - i.e. how many clusters to create)
  - e. Leveraging the clusters to efficiently select out a long list of potential locations for the cafe
  - f. Building and leveraging a location score (based on competing and complementary venues) to enable the selection of a short list

# Results & Discussion

Started with a grid of 162 candidate locations (600m apart from one another)

Leveraged the previously outlined methodology to filter down to the best possible locations - within a cluster of 41 locations which were central, vibrant, and with relatively fewer existing cafes. These are shown in blue, in the below visual:



# Results & Discussion

The long list was reduced to a 13 location short list, by leveraging the location score which was built. This allowed for a quantification of the location by trading off competing venues and complementary venues. The final locations for consideration are shown below



# Conclusion

Thus far, initial locations have been identified that are suitable, based on analysis of Foursquare data on the local venues, and accounting for the following criteria:

- Central
- Less competed
- More vibrant, and with complementary venues

The next steps will be as follows:

- Factor in available properties for the cafe, and overlay these with the existing recommendations
- Conduct the same analytical process with the actual candidate locations, and leverage the results to inform the final choice