Finding a location to a gourmet restaurant in Buenos Aires, Argentina

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Capstone Project - The Battle of the Neighborhoods

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Introduction

The Ciudad Autónoma de Buenos Aires (CABA) is one of the twenty-four federal entities and the capital city of Argentina. Buenos Aires is a cosmopolitan city and a major tourist destination. Its complex infrastructure makes it one of the most important metropolises in Latin America and is a global city of alpha, category given its influences on commerce, finance, fashion, art, gastronomy, education, entertainment and mainly in its marked culture.

In this project I am going to try to find an optimal location for a company that wants to open a gourmet restaurant in this city. The decision will be taken based on the places where we can find more restaurants, galleries, because those neighborhoods are supposed to by haunted.

In order to do that, first we are going to collect the data of the Buenos Aires neighborhoods and their respective coordinates. Then we will work with our database to finally apply K means clustering and thus determine in which neighborhoods the investment would be convenient. When we consider all these problems, we can create a map and information chart where the real estate index is placed on Buenos Aires and each district is clustered according to the venue density.

Data description

To consider the problem we can list the datas as below:

- I found the data from "Datos Abiertos de Buenos Aires" that contains a list of the different neighborhoods from CABA and their coordinates¹. I cleaned the data and convert it to a dataframe.
- I used Forsquare API to get the most common venues of given the neighborhoods from Buenos Aires.

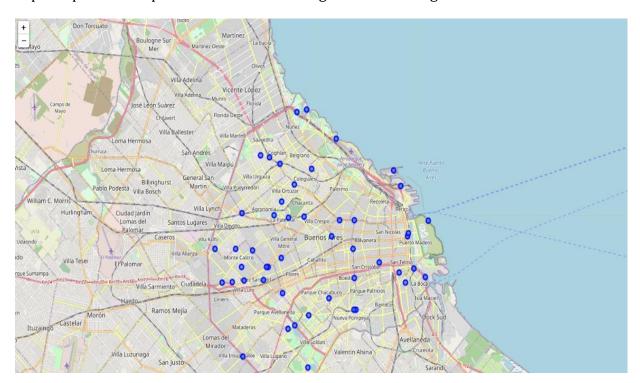
¹ http://cdn.buenosaires.gob.ar/datosabiertos/datasets/barrios/barrios.geojson

Methodology

As a database, I used GitHub repository in my study. My master data which has the main components Comuna, Neighborhoods, Latitude and Longitude information of the city.

| | comuna | Neighborhood | Latitude | Longitude |
|---|--------|------------------|------------|------------|
| 0 | 15 | CHACARITA | -34.595989 | -58.452820 |
| 1 | 15 | PATERNAL | -34.596558 | -58.465577 |
| 2 | 15 | VILLA CRESPO | -34.597827 | -58.423753 |
| 3 | 11 | VILLA DEL PARQUE | -34.614865 | -58.494610 |
| 4 | 5 | ALMAGRO | -34.614116 | -58.412870 |

I used python folium library to visualize geographic details of Buenos Aires and its neighborhoods and I created a map of Buenos Aires with neighborhoods superimposed on top. I used latitude and longitude values to get the visual as below:



I utilized the Foursquare API to explore the neighborhoods and segment them. I designed the limit as 100 venue and the radius 500 meter for each neighborhood from their given latitude and longitude information. Here is a head of the list Venues name, category, latitude, and longitude information from Forsquare API.

| | Neighborhood | Neighborhood Latitude | Neighborhood Longitude | Venue | Venue Latitude | Venue Longitude | Venue Category |
|---|--------------|--------------------------|---------------------------|---------------------|-------------------|--------------------|-------------------|
| 0 | CHACARITA | -34.595989 | -58.45282 | Yeite | -34.596012 | -58.449280 | Deli / Bodega |
| 1 | CHACARITA | -34.595989 | -58.45282 | Movistar Arena | -34.594348 | -58.448033 | Stadium |
| 2 | CHACARITA | -34.595989 | -58.45282 | Margen del Mundo | -34.596987 | -58.456835 | Museum |
| 3 | CHACARITA | -34.595989 | -58.45282 | Tiro Loco | -34.598935 | -58.452126 | Café |
| 4 | CHACARITA | -34.595989 | -58.45282 | Alumni Fútbol 5 | -34.597889 | -58.451936 | Soccer Field |

In summary, 146 unique categories were returned by Foursquare, then I created a table which shows list of top 10 venue category for each borough in below table.

| 1 ALMAG | hborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue | 8th Most Common Venue | 9th Most Common Venue |
|----------|----------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|------------------------------|-----------------------------|
| 2 BALVAI | ONOMIA | Hardware Store | Argentinian Restaurant | Train Station | Big Box Store | Bus Stop | Jewish Restaurant | Fast Food Restaurant | Furniture / Home Store | Food Truck |
| | AGRO | Argentinian Restaurant | Pizza Place | Indie Theater | Coffee Shop | Camera Store | Grocery Store | Restaurant | Pharmacy | Sushi Restaurant |
| 3 BARRA | /ANERA | Theater | Gym / Fitness Center | Hotel | Café | Italian Restaurant | Argentinian Restaurant | Gym | Pizza Place | Bakery |
| | RACAS | Supermarket | Gym | Restaurant | Café | Argentinian Restaurant | Pet Store | Plaza | Fish Market | Deli / Bodega |
| 4 BELGR | GRANO | Harbor / Marina | Argentinian Restaurant | Paintball Field | Sports Club | Gun Range | Women's Store | Fast Food Restaurant | Food Truck | Food Court |

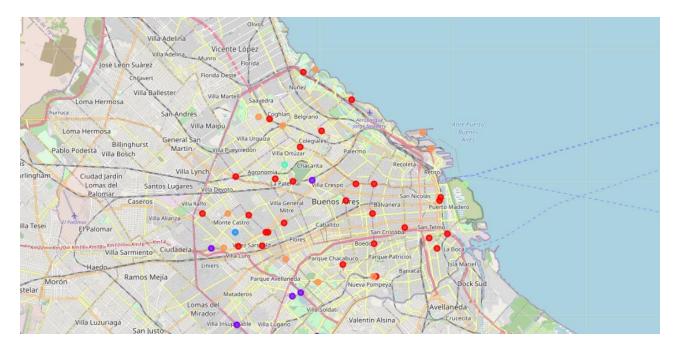
We have some common venue categories in neighborhoods. In this reason I used unsupervised learning K-means algorithm to cluster the neighborhoods. K-Means algorithm is one of the most common cluster methods of unsupervised learning.

I am going to run k-means to cluster the neighborhood into 6 clusters, and then, analyze the clusters looking for those with the greatest presence of soccer and sport venues

This is the new dataframe that includes the cluster as well as the top venues for each neighborhood.

| comuna | Neighborhood | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | | 5th Most Common Venue | 6th Mos Commo Venue |
|--------|---------------------|------------|------------|-------------------|-----------------------------|---------------------------------|-----------------------------|---------------------------|-----------------------------|---------------------------|
| 15 | CHACARITA | -34.595989 | -58.452820 | 1.0 | Museum | Bus Stop | Soccer Field | Stadium | Café | Deli / Bodega |
| 15 | PATERNAL | -34.596558 | -58.465577 | 0.0 | Art Gallery | BBQ Joint | Diner | Arts & Entertainment | Fish Market | Gastrop |
| 15 | VILLA CRESPO | -34.597827 | -58.423753 | 0.0 | Pizza Place | Middle Eastern Restaurant | Bar | BBQ Joint | Indie Theater | Café |
| 11 | VILLA DEL PARQUE | -34.614865 | -58.494610 | 0.0 | Breakfast Spot | BBQ Joint | Plaza | Argentinian Restaurant | Burger Joint | Restaur |
| 5 | ALMAGRO | -34.614116 | -58.412870 | 0.0 | Argentinian Restaurant | l | Indie Theater | Coffee Shop | Camera Store | Grocery Store |
| 4 | | | | | | | | | | . |

Finally, let's visualize the resulting clusters in a map of Buenos Aires.



Results

The findings suggest that neighborhoods in cluster number one, will be possible places to open the restaurant. That is because the results shown as that neighborhoods of that cluster have many Argentinian restaurants, which means that residents and turists visit those places and choose them to eating out.

| | Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue | 8th Most Common Venue |
|----|---------------------|-----------------------------|---------------------------------|-----------------------------|-----------------------------|-----------------------------|---------------------------------|---------------------------------|------------------------------|
| 1 | PATERNAL | Art Gallery | BBQ Joint | Diner | Arts & Entertainment | Fish Market | Gastropub | Gas Station | Furniture . Home Store |
| 2 | VILLA CRESPO | Pizza Place | Middle Eastern Restaurant | Bar | BBQ Joint | Indie Theater | Café | Cajun / Creole Restaurant | Restaurar |
| 3 | VILLA DEL PARQUE | Breakfast Spot | BBQ Joint | Plaza | Argentinian Restaurant | Burger Joint | Restaurant | Café | Fish Market |
| 4 | ALMAGRO | Argentinian Restaurant | Pizza Place | Indie Theater | Coffee Shop | Camera Store | Grocery Store | Restaurant | Pharmacy |
| 5 | CABALLITO | Café | Argentinian Restaurant | Bar | Indie Theater | Bakery | Hotel | Park | Restaurar |
| 6 | VILLA SANTA RITA | Restaurant | Middle Eastern Restaurant | Ice Cream Shop | Korean Restaurant | Deli / Bodega | Café | Furniture / Home Store | Comfort Food Restaurar |
| 8 | VILLA REAL | Pharmacy | Bus Stop | Café | Pizza Place | Argentinian Restaurant | Fast Food Restaurant | Breakfast Spot | Bakery |
| 10 | FLORESTA | Ice Cream Shop | Korean Restaurant | Deli / Bodega | Restaurant | Furniture / Home Store | Middle Eastern Restaurant | Comfort Food Restaurant | Toy / Game Store |

Neighborhoods in cluster number two seems to be places with little people movement, the most common venues are bus stop and soccer field, so I think that there is no place to a gourmet restaurant.

Clusters three, four and five although they are very tiny and have some restaurants, are discarded because comparing to cluster number one, they are not representative of restaurants. Same happens with cluster six, in which we can see some restaurants, but based on the results of our method, those neighborhoods seem to be more centrical, I mean, there are more pharmacies, gyms, etc.

Discussion

Although we have already given information to the stackholders about the optimal neighborhoods to establish the business, now it would be their job to find the location in one of these neighborhoods. With the use of data science tools, we could find the best place to rent according to cost and other variables of interest, but that is not part of this work.

Conclusion

The purpose of this project was to identify Buenos Aires neighborhoods that are optimal to invest in a gourmet restaurant. In order to do that first we collect the data of the Buenos Aires neighborhoods and their respective coordinates, we work the data and finally cluster the neighborhoods into 6 clusters. After a brief analysis we determined that cluster 1 was the most appropriate for our objective since there are lots of restaurants, and also some turistics attractions.