Analysing cultural activity of the neighbourhoods of Paris in order to chose the best place for starting a new restaurant

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Capstone Project - The Battle of Neighborhoods-

Introduction

- ✓ Paris is the most populous municipality and the capital of France.
- The city of Paris has 2.187 million inhabitants called Parisians. The Parisian agglomeration has 10.73 million inhabitants and its urban area has 12.78 million inhabitants. The Parisian agglomeration is thus the most populous in France, the third in Europe and the 32nd in the world.
- Paris symbolizes French culture and is nicknamed the City of Light, it is home to world famous monuments and is one of the most visited cities in the world.
- Paris has three opera houses, hundreds of theaters and cabarets, as well as a large number of music halls and performance venues.

Source: https://fr.wikipedia.org/wiki/Paris

Business problem and background

- Find the best place to open a new restaurant
- For this, we have to do a neighbourhood choice which would be probably the most likely to give a profitable business.
- Neighbourhood is one of choice inducteurs, like number of competitors and type of proposed services for instance, but probably than the level of cultural activity available in the neighbourhood, with the presence of several different venues close from the futur business, will help to develop it.
- Starting hypothesis: the more the cultural activity is available, the more competitors there are.

Description of the data

- ✓ Source for neighbourhood of Paris
 - Wikipedia page via data scraping:
 https://fr.wikipedia.org/wiki/Liste des quartiers administratifs de Paris
- Geographical coordinates of the neighbourhoods
 - Use of GeoPy library: https://geopy.readthedocs.io/en/stable/
 - ✓ And ArcGIS Online Geocoding Service: https://geocode.arcgis.com/arcgis/
- "Density" of cultural places and competing restaurants by neighbourhood
 - Use of Foursquare API: https://developer.foursquare.com/docs/places-api/

Methodology (1/2)

Selection of all venues by neighborhood / categories by neighborhood

```
Entrée [8]: def getNearbyVenues(names, latitudes, longitudes, radius=500):
            Function for getting all venues given in parameter "names"
            Return a DF with venue name, venue latitude & longitude, and its category (for each location name passed in parameters)
                venues list=[]
                for name, lat, lng in zip(names, latitudes, longitudes):
                    # create URL
                    url = 'https://api.foursquare.com/v2/venues/explore?&client id={}&client secret={}&v={}&ll={}, \
                          {}&radius={}&limit={}'.format(CLIENT ID, CLIENT SECRET, VERSION, lat, lng, radius, LIMIT)
                    # GET request
                    results = requests.get(url).json()["response"]['groups'][0]['items']
                    # return only relevant information for each nearby venue
                    venues list.append([(name, lat, lng, v['venue']['name'], v['venue']['location']['lat'], \
                                         v['venue']['location']['lng'], v['venue']['categories'][0]['name']) for v in results])
                nearby venues = pd.DataFrame([item for venue list in venues list for item in venue list])
                nearby venues.columns = ['Neighborhood', 'Neighborhood Latitude', 'Neighborhood Longitude', 'Venue', \
                                         'Venue Latitude', 'Venue Longitude', 'Venue Category']
                return(nearby venues)
```

Entrée []: paris_venues = getNearbyVenues(names=df['Neighborhood'], latitudes=df['Latitude'],longitudes=df['Longitude'])

Methodology (2/2)

One hot encoding

```
Entrée [53]: # one hot encoding
    paris_onehot = pd.get_dummies(paris_venues[['Venue Category']], prefix="", prefix_sep="")

# add neighborhood column back to dataframe
    paris_onehot['Neighborhood'] = paris_venues['Neighborhood']

# move neighborhood column to the first column
    fixed_columns = [paris_onehot.columns[-1]] + list(paris_onehot.columns[:-1])
    paris_onehot = paris_onehot[fixed_columns]
```

K-Means Clustering

```
Entrée [74]: # set number of clusters
kclusters = 8

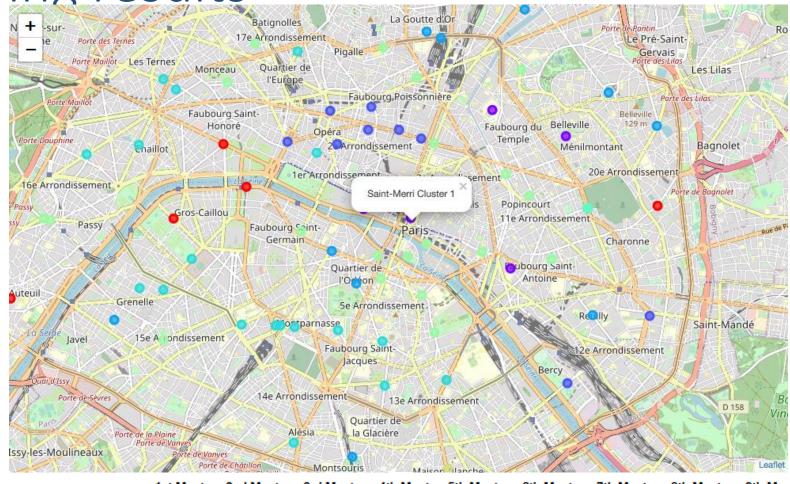
paris_grouped_clustering = paris_grouped.drop('Neighborhood', 1)

# run k-means clustering
kmeans = KMeans(n_clusters=kclusters, random_state=0).fit(paris_grouped_clustering)
```

Plotting on map with neighborhood coordinates

```
Entrée [7]: # create map of Paris using latitude and longitude values
                map paris = folium.Map(location=[latitude, longitude], zoom start=12)
                for lat, lng, bor, neigh in zip(df['Latitude'], df['Longitude'], df['Borough'], df['Neighborhood']):
                     label = '{} {}'.format(bor, neigh)
                    label = folium.Popup(label, parse html=True)
                     folium.CircleMarker([lat, lng], radius=5, popup=label, color='blue',fill=True,
                                           fill color='#3186cc', fill opacity=0.7, parse html=False).add to(map paris)
                map paris
                        Colombes
                                                                                                         Drancy
                                                                            Aubervilliers
                                                         Saint-Ouen-
                                     Asnières-sur-
                                                                                                                              Les Pavillons-
                                                          sur-Seine
                      Garenne
                                                                                                        Bobigny
                                                                                                                                sous-Bois
                     Colombes
                                               Clichy
                                                                                                                                      Le Raino
                        Courbevoie
                                                              Clignancour
                                    Levallois-Perret
     Nanterre
                                                                                                                         a Mare à la
                                                           e Arrondissement
                                                                                                         Noisy-le-Sec
                                                                                                   Romainville
                                                                                     Le Pré-Saint-
                  Puteaux
                                                                                                                                 Villemomble
                                                                                        Gervais
naison
                                                                                                                         Rosny-sous-
              Suresnes
                                                                                    montant Bagnolet
                                                                                                      Montreuil
                                                                                                                             Neuilly-Plaisance
uzenval
                                                                                                                  Fontenay-sous-
                                                                                                    -Vincennes
                                                                                  Reolly Saint-Mandé
         Saint-Cloud
                                                                                                                                 Le Perreux
                                                                                                                                  sur-Marne
            Boulogne-Billancourt
                                                             Ouartier de
-d'Avray
                            Issy-les-Moulineaux
        Sèvres
                                                                                           Charenton-
                                                                                                                Joinville-le-
                                                   Montrouge
                                                                                                                                  Champigny
                                                                   Le Kremlin-
                  Meudon
                                                                              Ivry-sur-Seine
                                                                                                                                   sur-Marne
                                                                    Bicêtre
                                                                                                                        Saint-Maur
                                                                                                                        des-Fossés
                                                                                                Maisons-Alfort
                             Clamart Châtillon
                                             Bagneux
```

Clustering results



	Neighborhood	Longitude	Cluster Labels	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	10th Most Common Venue
12	Saint-Merri	2.350530	1	Art Gallery	French Restaurant	Plaza	Restaurant	Ice Cream Shop	Art Museum	Park	Lebanese Restaurant	Pub	Burger Joint

Discussion

- ✓ The most suitable neighborhood with cultural activities are mainly in cluster 2, 3, 4 et 5.
- The cluster 5, with Theater venues, could a good choice for starting an evening restaurant.
- Clusters 3 and 4 are place for Museums and Historic sites, so more for day restaurant, but the competitors are already numerous.
- The cluster 2 has a large variety of cultural activities (Museums, Art galleries, Historic sites, Exhibits, Theatres) so a possibility to start a restaurant with spread opened hours on all day. And concurrence seems to be less important.
- The best neighborhood could be "St Merri" to open a new restaurant ou food business.

Conclusion

- BeautifulSoup, GeoPy and Folium are very useful Python libraries which are quite easy to use and give a real avantage to analyse geographical data.
- Data analysis and machine learning can be a help to deals with some complex business problems.
- The K-means algorithm, for the unsupervised learning, has help in our case to cluster the similar Paris neighborhoods.
- Finally, we can give a recommendation and advice (and the better places) for who want to launch a business such a new restaurant.