# **Capstone Project - The Battle of Neighborhoods (Week 2)**

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# Introduction: Business Problem

Due to the **Covid-19 crisis**, many businesses have had to close. I have been able to appreciate this horrible situation especially in the neighborhood where I live.

When the novel coronavirus pandemic initially hit back in March and politicians ordered all restaurants and bars to close to in-room dining, a panic soared throughout the hospitality industry. Thousands of establishments shut their doors immediately for indoor dining, leaving countless people out of work. Many questions arose: When could restaurants fully reopen? How many might not return? What would the landscape look like with so much restaurant vacancy?

I'm going to analyze the number of food businesses in the area where I live to see if it would be a good business opportunity to open a restaurant at this moment. It is a good way to revive the economy and bring the neighborhood back to life.

## **Data**

I am going to check the restaurants in my neighborhoods and analyze the most surrounding areas to conclude if there are business opportunities for food places at the moment.

Based on definition of our problem, factors that will influence our decission are:

- number of existing restaurants in the neighborhood (any type of restaurant)
- · number of and distance to different sports centers in the neighborhood, if any

I will import the coordinates of my neighborhood and visualize using **Foursquare API** the restaurants and bars that are nearby. Also,I wil show the different sport centers as it is very typical to have a drink after doing sports with friends. I'm gonna use Folium for the graphics.

Import necessary libraries

```
In [2]:
        import requests # library to handle requests
        import pandas as pd # library for data analsysis
        import numpy as np # library to handle data in a vectorized manner
        import random # library for random number generation
        !conda install -c conda-forge geopy --yes
        from geopy.geocoders import Nominatim # module to convert an address into latitude and longitude values
        # libraries for displaying images
        from IPython.display import Image
        from IPython.core.display import HTML
        # tranforming json file into a pandas dataframe library
        from pandas.io.json import json_normalize
        !conda install -c conda-forge folium=0.5.0 --yes
        import folium # plotting library
        print('Folium installed')
        print('Libraries imported.')
        Requirement already satisfied: geopy in /opt/conda/envs/Python36/lib/python3.6/site-packages (2.0.0)
        Requirement already satisfied: geographiclib<2,>=1.49 in /opt/conda/envs/Python36/lib/python3.6/site-packag
        es (from geopy) (1.50)
        Solving environment: done
        # All requested packages already installed.
        Solving environment: done
        # All requested packages already installed.
        Libraries imported.
```

#### **Define Foursquare Credentials and Version**

```
In [351]: # @hidden_cell
CLIENT_ID = 'ET2ZUTRFK1440YYWTR3VSKMMIZBHTUHN10WOMR5OACTNX05Y' # your Foursquare ID
CLIENT_SECRET = 'EBLM42LFLGGPLXXGBALASOP2SKXNFPBUHRWIGY1LM4SAZQ5A' # your Foursquare Secret
VERSION = '20180604'
LIMIT = 50
```

### Let's visualize the restaurants that are nearby

The geograpical coordinates of my neighborhood are 40.4846746, -3.7218096.

```
In [353]: map_lacoma = folium.Map(location=[latitude, longitude], zoom_start=15)
    folium.Marker([latitude, longitude], popup='Point of interest').add_to(map_lacoma)
    map_lacoma
```

Out[353]: Make this Notebook Trusted to load map: File -> Trust Notebook

```
In [354]: radius = 1000
    food_category = '4d4b7105d754a06374d81259'
    url = 'https://api.foursquare.com/v2/venues/explore?client_id={}&client_secret={}&ll={},{}&v={}&radius={}&li
    mit={}'.format(CLIENT_ID, CLIENT_SECRET, latitude, longitude, VERSION, radius, LIMIT)
    import requests
    results = requests.get(url).json()
    items = results['response']['groups'][0]['items']
```

	name	categories	address	СС	city	country	crossStreet	distance	formatted Address	labeledLatLngs	
0	El Viejo Almacén de Buenos Aires	Argentinian Restaurant	C. Ramón Gómez de la Serna, 4	ES	Madrid	España	NaN	416	[C. Ramón Gómez de la Serna, 4, 28035 Madrid M	[{'label': 'display', 'lat': 40.48112446689104	40.4811
1	Tras Os Montes	Seafood Restaurant	Senda del Infante 28	ES	Madrid	España	NaN	730	[Senda del Infante 28, 28049 Madrid Madrid, Es	[{'label': 'display', 'lat': 40.49122842350793	40.4912
2	La Alternativa	Brewery	NaN	ES	Madrid Centro, Madrid	España	NaN	366	[Madrid Centro Madrid, España]	[{'label': 'display', 'lat': 40.48678467998688	40.4867
3	Akira Comics	Comic Shop	Av de Betanzos, 74	ES	Madrid	España	Fermín Caballero	725	[Av de Betanzos, 74 (Fermín Caballero), 28034 	[{'label': 'display', 'lat': 40.48210086100509	40.4821
4	CDM Go Fit Peñagrande	Gym / Fitness Center	Avenida Monforte de Lemos, 184	ES	Madrid	España	NaN	923	[Avenida Monforte de Lemos, 184, 28035 Madrid	[{'label': 'display', 'lat': 40.47706381409571	40.4770
5	La Maso Sport Center - Padel	Sports Club	Calle de la Masó, 34	ES	Madrid	España	NaN	863	[Calle de la Masó, 34, 28034 Madrid Madrid, Es	[{'label': 'display', 'lat': 40.489904, 'lng':	40.4899
6	Singapur	Chinese Restaurant	C. Mirador de la Reina, 99	ES	Madrid	España	NaN	555	[C. Mirador de la Reina, 99, 28035 Madrid Madr	[{'label': 'display', 'lat': 40.48914619140871	40.4891
7	Los Granadinos	Tapas Restaurant	Av del Cardenal Herrera Oria, 248	ES	Madrid	España	NaN	402	[Av del Cardenal Herrera Oria, 248, 28035 Madr	[{'label': 'display', 'lat': 40.48112023451070	40.4811
8	Cala Millor	Bakery	C. Fermín Caballero, 70	ES	Madrid	España	Av. Betanzos	861	[C. Fermín Caballero, 70 (Av. Betanzos), 28034	[{'label': 'display', 'lat': 40.48215283169287	40.4821
9	Body Factory Mirasierra	Gym	C/ Valle de en medio 15	ES	Madrid	España	NaN	548	[C/ Valle de en medio 15, Madrid Madrid, España]	[{'label': 'display', 'lat': 40.48838053202037	40.4883
10	The Clover Irish Tavern	Pub	C. Valle de Pinares Llanos, 8	ES	Madrid	España	NaN	303	[C. Valle de Pinares Llanos, 8, 28035 Madrid M	[{'label': 'display', 'lat': 40.4858706510018,	40.4858
11	Churreria Avenida	Café	C. Cerro Minguete, 16	ES	Madrid	España	NaN	356	[C. Cerro Minguete, 16, 28035 Madrid Madrid, E	[{'label': 'display', 'lat': 40.487003559763,	40.4870
12	Mesón Las Vegas	Spanish Restaurant	Calle de la Isla de Arosa, 29, 28035 Madrid	ES	Madrid	España	NaN	650	[Calle de la Isla de Arosa, 29, 28035 Madrid,	[{'label': 'display', 'lat': 40.48034557270606	40.4803
13	Ginos Mirasierra	Italian Restaurant	C. del Mirador de la Reina, 107	ES	Madrid	España	NaN	660	[C. del Mirador de la Reina, 107, 28035 Madrid	[{'label': 'display', 'lat': 40.49059779325799	40.4905
14	Peggy Sue's	Diner	Av. Monforte de Lemos 185	ES	Madrid	España	NaN	960	[Av. Monforte de Lemos 185, 28035 Madrid Madri	[{'label': 'display', 'lat': 40.47651545133406	40.4765
15	Supercor Express	Food & Drink Shop	C/ Cerro Minguete, 9	ES	Madrid	España	NaN	308	[C/ Cerro Minguete, 9, 28035 Madrid Madrid, Es	[{'label': 'display', 'lat': 40.48700313082754	40.4870
16	Bonsai	Chinese Restaurant	Valle Pinares Llanos, 8	ES	Madrid	España	NaN	330	[Valle Pinares Llanos, 8, 28035 Madrid Madrid,	[{'label': 'display', 'lat': 40.48579063622070	40.4857
17	Lazcano	Café	NaN	ES	NaN	España	NaN	601	[España]	[{'label': 'display', 'lat': 40.48818724029966	40.4881
18	La Ponderosa Bar	Tapas Restaurant	Av. Cardenal Herrera Oria, 231	ES	Madrid	España	NaN	459	[Av. Cardenal Herrera Oria, 231, 28034 Madrid	[{'label': 'display', 'lat': 40.48212002479462	40.4821
19	Los Granadinos	Bar	C. Leopoldo Alas Clarín, 53	ES	Madrid	España	NaN	74	[C. Leopoldo Alas Clarín, 53, 28035 Madrid Mad	[{'label': 'display', 'lat': 40.48454075118334	40.4845
20	Polideportivo La Masó	Athletics & Sports	C. La Masó, 80	ES	Madrid	España	Av. Ventisquero de la Condesa	546	[C. La Masó, 80 (Av. Ventisquero de la Condesa	[{'label': 'display', 'lat': 40.48450797800188	40.4845

	name	categories	address	СС	city	country	crossStreet	distance	formattedAddress	labeledLatLngs	
21	SuperCor Exprés	Convenience Store	NaN	ES	NaN	España	NaN	985	[España]	[{'label': 'display', 'lat': 40.4759056665114,	40.4759
22	Cervecería Granada	Bar	NaN	ES	Madrid	España	NaN	131	[Madrid Madrid, España]	[{'label': 'display', 'lat': 40.48474093558645	40.4847
23	Campo de Fútbol de Lacoma	BBQ Joint	Calle Ramón Gómez de la Serna	ES	Madrid	España	NaN	139	[Calle Ramón Gómez de la Serna, Madrid Madrid,	[{'label': 'display', 'lat': 40.48484263040647	40.4848
24	Polideportivo Isla Tabarca	Soccer Field	NaN	ES	NaN	España	NaN	871	[España]	[{'label': 'display', 'lat': 40.47687381375908	40.4768
25	El Mirador	Restaurant	Calle Mirador de la Reina 99	ES	Madrid	España	Cerro de Álamos Blancos	543	[Calle Mirador de la Reina 99 (Cerro de Álamos	[{'label': 'display', 'lat': 40.489086, 'lng':	40.4890
26	VIPS Mirasierra	Burger Joint	C. Mirador de la Reina, 107	ES	Madrid	España	NaN	663	[C. Mirador de la Reina, 107, 28035 Madrid Mad	[{'label': 'display', 'lat': 40.49061956122534	40.4906
27	Instalación Deportiva Valdeyeros	Soccer Field	Rosalía de Castro	ES	Madrid	España	NaN	725	[Rosalía de Castro, Madrid Madrid, España]	[{'label': 'display', 'lat': 40.48385972087871	40.4838
28	Starving Conciertos	Bar	C. Fermin Caballero, 69	ES	Madrid	España	NaN	937	[C. Fermin Caballero, 69, 28029 Madrid Madrid,	[{'label': 'display', 'lat': 40.48199789977096	40.4819
29	Ginos Monforte de Lemos	Italian Restaurant	Av. Monforte de Lemos, 193	ES	Madrid	España	NaN	980	[Av. Monforte de Lemos, 193, 28035 Madrid Madr	[{'label': 'display', 'lat': 40.47600665515477	40.4760
30	Bar Carmelo	Breakfast Spot	Isla de Saipán 73	ES	Madrid	España	NaN	923	[Isla de Saipán 73, 28035 Madrid Madrid, España]	[{'label': 'display', 'lat': 40.47923696814644	40.4792
31	Gama	Grocery Store	La maso 2	ES	Madrid	España	NaN	965	[La maso 2, Madrid Madrid, España]	[{'label': 'display', 'lat': 40.48176480591085	40.4817
32	Restaurante Tres Carro's	Restaurant	Alfonso Rodríguez Castelao	ES	NaN	España	NaN	983	[Alfonso Rodríguez Castelao, España]	[{'label': 'display', 'lat': 40.48205674009662	40.4820

```
In [356]:
                               #Cleaning
                               dataframe_filtered.drop(dataframe_filtered[ (dataframe_filtered['categories'] == 'Comic Shop') ].index , inp
                               lace=True)
                               {\tt dataframe\_filtered.(dataframe\_filtered['categories'] == "Bakery') \ ]. index \ , \ inplace \ .}
                               \label{lem:dataframe_filtered} \texttt{dataframe\_filtered['categories'] == 'Cafe') } ]. index \ , \ inplace = Tafe' | Tafe
                               rue)
                               dataframe_filtered.drop(dataframe_filtered[ (dataframe_filtered['categories'] == 'Food & Drink Shop') ].inde
                               x , inplace=True)
                               dataframe_filtered.drop(dataframe_filtered[ (dataframe_filtered['categories'] == 'Convenience Store') ].inde
                               x , inplace=True)
                               dataframe_filtered.drop(dataframe_filtered[ (dataframe_filtered['categories'] == 'BBQ Joint') ].index , inpl
                               dataframe_filtered.drop(dataframe_filtered[ (dataframe_filtered['categories'] == 'Soccer Field') ].index , i
                               nplace=True)
                               dataframe_filtered.drop(dataframe_filtered[ (dataframe_filtered['categories'] == 'Grocery Store') ].index ,
                               inplace=True)
                               dataframe_filtered.reset_index(inplace=True, drop=True)
```

dataframe\_filtered.head(32)

	name	categories	address	сс	city	country	crossStreet	distance	formattedAddress	labeledLatLngs	la
0	El Viejo Almacén de Buenos Aires	Argentinian Restaurant	C. Ramón Gómez de la Serna, 4	ES	Madrid	España	NaN	416	[C. Ramón Gómez de la Serna, 4, 28035 Madrid M	[{'label': 'display', 'lat': 40.48112446689104	40.48112
1	Tras Os Montes	Seafood Restaurant	Senda del Infante 28	ES	Madrid	España	NaN	730	[Senda del Infante 28, 28049 Madrid Madrid, Es	[{'label': 'display', 'lat': 40.49122842350793	40.49122
2	La Alternativa	Brewery	NaN	ES	Madrid Centro, Madrid	España	NaN	366	[Madrid Centro Madrid, España]	[{'label': 'display', 'lat': 40.48678467998688	40.48678
3	CDM Go Fit Peñagrande	Gym / Fitness Center	Avenida Monforte de Lemos, 184	ES	Madrid	España	NaN	923	[Avenida Monforte de Lemos, 184, 28035 Madrid	[{'label': 'display', 'lat': 40.47706381409571	40.477064
4	La Maso Sport Center - Padel	Sports Club	Calle de la Masó, 34	ES	Madrid	España	NaN	863	[Calle de la Masó, 34, 28034 Madrid Madrid, Es	[{'label': 'display', 'lat': 40.489904, 'lng':	40.489904
5	Singapur	Chinese Restaurant	C. Mirador de la Reina, 99	ES	Madrid	España	NaN	555	[C. Mirador de la Reina, 99, 28035 Madrid Madr	[{'label': 'display', 'lat': 40.48914619140871	40.48914
6	Los Granadinos	Tapas Restaurant	Av del Cardenal Herrera Oria, 248	ES	Madrid	España	NaN	402	[Av del Cardenal Herrera Oria, 248, 28035 Madr	[{'label': 'display', 'lat': 40.48112023451070	40.48112
7	Body Factory Mirasierra	Gym	C/ Valle de en medio 15	ES	Madrid	España	NaN	548	[C/ Valle de en medio 15, Madrid Madrid, España]	[{'label': 'display', 'lat': 40.48838053202037	40.48838
8	The Clover Irish Tavern	Pub	C. Valle de Pinares Llanos, 8	ES	Madrid	España	NaN	303	[C. Valle de Pinares Llanos, 8, 28035 Madrid M	[{'label': 'display', 'lat': 40.4858706510018,	40.48587
9	Mesón Las Vegas	Spanish Restaurant	Calle de la Isla de Arosa, 29, 28035 Madrid	ES	Madrid	España	NaN	650	[Calle de la Isla de Arosa, 29, 28035 Madrid,	[{'label': 'display', 'lat': 40.48034557270606	40.48034
10	Ginos Mirasierra	Italian Restaurant	C. del Mirador de la Reina, 107	ES	Madrid	España	NaN	660	[C. del Mirador de la Reina, 107, 28035 Madrid	[{'label': 'display', 'lat': 40.49059779325799	40.49059
11	Peggy Sue's	Diner	Av. Monforte de Lemos 185	ES	Madrid	España	NaN	960	[Av. Monforte de Lemos 185, 28035 Madrid Madri	[{'label': 'display', 'lat': 40.47651545133406	40.47651
12	Bonsai	Chinese Restaurant	Valle Pinares Llanos, 8	ES	Madrid	España	NaN	330	[Valle Pinares Llanos, 8, 28035 Madrid Madrid,	[{'label': 'display', 'lat': 40.48579063622070	40.48579
13	La Ponderosa Bar	Tapas Restaurant	Av. Cardenal Herrera Oria, 231	ES	Madrid	España	NaN	459	[Av. Cardenal Herrera Oria, 231, 28034 Madrid	[{'label': 'display', 'lat': 40.48212002479462	40.482120
14	Los Granadinos	Bar	C. Leopoldo Alas Clarín, 53	ES	Madrid	España	NaN	74	[C. Leopoldo Alas Clarín, 53, 28035 Madrid Mad	[{'label': 'display', 'lat': 40.48454075118334	40.48454
15	Polideportivo La Masó	Athletics & Sports	C. La Masó, 80	ES	Madrid	España	Av. Ventisquero de la Condesa	546	[C. La Masó, 80 (Av. Ventisquero de la Condesa	[{'label': 'display', 'lat': 40.48450797800188	40.48450
16	Cervecería Granada	Bar	NaN	ES	Madrid	España	NaN	131	[Madrid Madrid, España]	[{'label': 'display', 'lat': 40.48474093558645	40.48474
17	El Mirador	Restaurant	Calle Mirador de la Reina 99	ES	Madrid	España	Cerro de Álamos Blancos	543	[Calle Mirador de la Reina 99 (Cerro de Álamos	[{'label': 'display', 'lat': 40.489086, 'lng':	40.489080
18	VIPS Mirasierra	Burger Joint	C. Mirador de la Reina, 107	ES	Madrid	España	NaN	663	[C. Mirador de la Reina, 107, 28035 Madrid Mad	[{'label': 'display', 'lat': 40.49061956122534	40.490620
19	Starving Conciertos	Bar	C. Fermin Caballero, 69	ES	Madrid	España	NaN	937	[C. Fermin Caballero, 69, 28029 Madrid Madrid,	[{'label': 'display', 'lat': 40.48199789977096	40.48199

	name	categories	address	СС	city	country	crossStreet	distance	formatted Address	labeledLatLngs	la
20	Ginos Monforte de Lemos	Italian Restaurant	Av. Monforte de Lemos, 193	ES	Madrid	España	NaN	980	[Av. Monforte de Lemos, 193, 28035 Madrid Madr	[{'label': 'display', 'lat': 40.47600665515477	40.47600°
21	Bar Carmelo	Breakfast Spot	Isla de Saipán 73	ES	Madrid	España	NaN	923	[Isla de Saipán 73, 28035 Madrid Madrid, España]	[{'label': 'display', 'lat': 40.47923696814644	40.47923
22	Restaurante Tres Carro's	Restaurant	Alfonso Rodríguez Castelao	ES	NaN	España	NaN	983	[Alfonso Rodríguez Castelao, España]	[{'label': 'display', 'lat': 40.48205674009662	40.48205

```
In [357]:
          #Bar
          dataframe_filtered_bar = dataframe_filtered.iloc[0:0,:].copy()
          cond = (dataframe_filtered['categories'] == 'Brewery')
          rows = dataframe_filtered.loc[cond, :]
          dataframe_filtered_bar = dataframe_filtered_bar.append(rows, ignore_index=True)
          dataframe_filtered.drop(rows.index, inplace=True)
          dataframe_filtered_bar.drop(dataframe_filtered_bar.index, inplace=True)
          cond = (dataframe_filtered['categories'] == 'Bar')
          rows = dataframe_filtered.loc[cond, :]
          dataframe filtered bar = dataframe filtered bar.append(rows, ignore index=True)
          dataframe_filtered.drop(rows.index, inplace=True)
          cond = (dataframe_filtered['categories'] == 'Breakfast Spot')
          rows = dataframe_filtered.loc[cond, :]
          dataframe_filtered_bar = dataframe_filtered_bar.append(rows, ignore_index=True)
          dataframe_filtered.drop(rows.index, inplace=True)
          cond = (dataframe filtered['categories'] == 'Snack Place')
          rows = dataframe_filtered.loc[cond, :]
          dataframe_filtered_bar = dataframe_filtered_bar.append(rows, ignore_index=True)
          dataframe_filtered.drop(rows.index, inplace=True)
          cond = (dataframe_filtered['categories'] == 'Tapas Restaurant')
          rows = dataframe_filtered.loc[cond, :]
          dataframe_filtered_bar = dataframe_filtered_bar.append(rows, ignore_index=True)
          dataframe_filtered.drop(rows.index, inplace=True)
          cond = (dataframe_filtered['categories'] == 'Pub')
          rows = dataframe_filtered.loc[cond, :]
          dataframe_filtered_bar = dataframe_filtered_bar.append(rows, ignore_index=True)
          dataframe_filtered.drop(rows.index, inplace=True)
          dataframe_filtered_bar.reset_index(inplace=True, drop=True)
          dataframe_filtered_bar.head(40)
```

#### Out[357]:

	name	categories	address	СС	city	country	crossStreet	distance	formattedAddress	labeledLatLngs	lat
0	Los Granadinos	Bar	C. Leopoldo Alas Clarín, 53	ES	Madrid	España	NaN	74	[C. Leopoldo Alas Clarín, 53, 28035 Madrid Mad	[{'label': 'display', 'lat': 40.48454075118334	40.484541
1	Cervecería Granada	Bar	NaN	ES	Madrid	España	NaN	131	[Madrid Madrid, España]	[{'label': 'display', 'lat': 40.48474093558645	40.484741
2	Starving Conciertos	Bar	C. Fermin Caballero, 69	ES	Madrid	España	NaN	937	[C. Fermin Caballero, 69, 28029 Madrid Madrid,	[{'label': 'display', 'lat': 40.48199789977096	40.481998
3	Bar Carmelo	Breakfast Spot	Isla de Saipán 73	ES	Madrid	España	NaN	923	[Isla de Saipán 73, 28035 Madrid Madrid, España]	[{'label': 'display', 'lat': 40.47923696814644	40.479237
4	Los Granadinos	Tapas Restaurant	Av del Cardenal Herrera Oria, 248	ES	Madrid	España	NaN	402	[Av del Cardenal Herrera Oria, 248, 28035 Madr	[{'label': 'display', 'lat': 40.48112023451070	40.481120
5	La Ponderosa Bar	Tapas Restaurant	Av. Cardenal Herrera Oria, 231	ES	Madrid	España	NaN	459	[Av. Cardenal Herrera Oria, 231, 28034 Madrid	[{'label': 'display', 'lat': 40.48212002479462	40.482120
6	The Clover Irish Tavern	Pub	C. Valle de Pinares Llanos, 8	ES	Madrid	España	NaN	303	[C. Valle de Pinares Llanos, 8, 28035 Madrid M	[{'label': 'display', 'lat': 40.4858706510018,	40.485871

```
In [358]:
           #Sports
           dataframe_filtered_sport = dataframe_filtered.iloc[0:0,:].copy()
           cond = (dataframe_filtered['categories'] == 'Gym / Fitness Center')
           rows = dataframe_filtered.loc[cond, :]
           dataframe_filtered_sport = dataframe_filtered_sport.append(rows, ignore_index=True)
dataframe_filtered.drop(rows.index, inplace=True)
           cond = (dataframe_filtered['categories'] == 'Sports Club')
           rows = dataframe_filtered.loc[cond, :]
           dataframe_filtered_sport = dataframe_filtered_sport.append(rows, ignore_index=True)
           dataframe_filtered.drop(rows.index, inplace=True)
           cond = (dataframe_filtered['categories'] == 'Gym')
           rows = dataframe_filtered.loc[cond, :]
           dataframe_filtered_sport = dataframe_filtered_sport.append(rows, ignore_index=True)
           dataframe_filtered.drop(rows.index, inplace=True)
           cond = (dataframe_filtered['categories'] == 'Athletics & Sports')
           rows = dataframe_filtered.loc[cond, :]
           dataframe_filtered_sport = dataframe_filtered_sport.append(rows, ignore_index=True)
           dataframe_filtered.drop(rows.index, inplace=True)
           {\tt dataframe\_filtered\_sport.reset\_index(inplace=True, \; drop=True)}
           dataframe_filtered_sport.head(40)
```

#### Out[358]:

	name	categories	address	СС	city	country	crossStreet	distance	formattedAddress	labeledLatLngs	lat
0	CDM Go Fit Peñagrande	Gym / Fitness Center	Avenida Monforte de Lemos, 184	ES	Madrid	España	NaN	923	[Avenida Monforte de Lemos, 184, 28035 Madrid	[{'label': 'display', 'lat': 40.47706381409571	40.477064
1	La Maso Sport Center - Padel	Sports Club	Calle de la Masó, 34	ES	Madrid	España	NaN	863	[Calle de la Masó, 34, 28034 Madrid Madrid, Es	[{'label': 'display', 'lat': 40.489904, 'lng':	40.489904
2	Body Factory Mirasierra	Gym	C/ Valle de en medio 15	ES	Madrid	España	NaN	548	[C/ Valle de en medio 15, Madrid Madrid, España]	[{'label': 'display', 'lat': 40.48838053202037	40.488381
3	Polideportivo La Masó	Athletics & Sports	C. La Masó, 80	ES	Madrid	España	Av. Ventisquero de la Condesa	546	[C. La Masó, 80 (Av. Ventisquero de la Condesa	[{'label': 'display', 'lat': 40.48450797800188	40.484508

In [359]: dataframe\_filtered.reset\_index(inplace=True, drop=True) dataframe\_filtered.head(40)

Out[359]:

	name	categories	address	СС	city	country	crossStreet	distance	formattedAddress	labeledLatLngs	lat
0	El Viejo Almacén de Buenos Aires	Argentinian Restaurant	C. Ramón Gómez de Ia Serna, 4	ES	Madrid	España	NaN	416	[C. Ramón Gómez de la Serna, 4, 28035 Madrid M	[{'label': 'display', 'lat': 40.48112446689104	40.481124
1	Tras Os Montes	Seafood Restaurant	Senda del Infante 28	ES	Madrid	España	NaN	730	[Senda del Infante 28, 28049 Madrid Madrid, Es	[{'label': 'display', 'lat': 40.49122842350793	40.491228
2	Singapur	Chinese Restaurant	C. Mirador de la Reina, 99	ES	Madrid	España	NaN	555	[C. Mirador de la Reina, 99, 28035 Madrid Madr	[{'label': 'display', 'lat': 40.48914619140871	40.489146
3	Mesón Las Vegas	Spanish Restaurant	Calle de la Isla de Arosa, 29, 28035 Madrid	ES	Madrid	España	NaN	650	[Calle de la Isla de Arosa, 29, 28035 Madrid,	[{'label': 'display', 'lat': 40.48034557270606	40.480346
4	Ginos Mirasierra	Italian Restaurant	C. del Mirador de la Reina, 107	ES	Madrid	España	NaN	660	[C. del Mirador de la Reina, 107, 28035 Madrid	[{'label': 'display', 'lat': 40.49059779325799	40.490598
5	Peggy Sue's	Diner	Av. Monforte de Lemos 185	ES	Madrid	España	NaN	960	[Av. Monforte de Lemos 185, 28035 Madrid Madri	[{'label': 'display', 'lat': 40.47651545133406	40.476515
6	Bonsai	Chinese Restaurant	Valle Pinares Llanos, 8	ES	Madrid	España	NaN	330	[Valle Pinares Llanos, 8, 28035 Madrid Madrid,	[{'label': 'display', 'lat': 40.48579063622070	40.485791
7	El Mirador	Restaurant	Calle Mirador de la Reina 99	ES	Madrid	España	Cerro de Álamos Blancos	543	[Calle Mirador de la Reina 99 (Cerro de Álamos	[{'label': 'display', 'lat': 40.489086, 'lng':	40.489086
8	VIPS Mirasierra	Burger Joint	C. Mirador de la Reina, 107	ES	Madrid	España	NaN	663	[C. Mirador de la Reina, 107, 28035 Madrid Mad	[{'label': 'display', 'lat': 40.49061956122534	40.490620
9	Ginos Monforte de Lemos	Italian Restaurant	Av. Monforte de Lemos, 193	ES	Madrid	España	NaN	980	[Av. Monforte de Lemos, 193, 28035 Madrid Madr	[{'label': 'display', 'lat': 40.47600665515477	40.476007
10	Restaurante Tres Carro's	Restaurant	Alfonso Rodríguez Castelao	ES	NaN	España	NaN	983	[Alfonso Rodríguez Castelao, España]	[{'label': 'display', 'lat': 40.48205674009662	40.482057

# Methodology

In this project I'm trying to detect the areas of my neighborhood that have low restaurant density. I will limit my analysis to a radius of ~1km around the metro station.

In the first step, I have collected the required data: location and type (category) of every restaurant within 1km from the metro station of the neighborhood (Lacoma). I have also identified restaurants nearby (according to Foursquare categorization).

In the second step in my analysis, I will be calculating and explorating the different restaurant radius to identify a few promising areas close to the metro station with low number of restaurants in general, and focus our attention on those areas.

In the third and final step, I will analyze different sport services that are nearby the selected area to complete de studio. In the neighborhood, it is very typical to have a drink or have dinner around sports activities.

# **Analysis**

Let's take a look at the restaurants and pubs in the area.

```
In [360]:
           venues_map = folium.Map(location=[latitude, longitude], zoom_start=15) # generate map centred around the met
           ro station
           # add a red circle marker to represent the metro station
           folium.features.CircleMarker(
                [latitude, longitude],
                radius=10,
                color='black',
                popup='Metro Station',
                fill = True,
                fill_color = 'black',
                fill_opacity = 0.6
           ).add_to(venues_map)
           for lat, lng, label in zip(dataframe_filtered.lat, dataframe_filtered.lng, dataframe_filtered.categories):
                folium.features.CircleMarker(
                    [lat, lng],
                    radius=5,
                    color='blue',
                    popup=label,
                    fill = True,
                    fill_color='blue',
                    fill_opacity=0.6
                ).add_to(venues_map)
           for lat, lng, label in zip(dataframe_filtered_bar.lat, dataframe_filtered_bar.lng, dataframe_filtered_bar.ca
                folium.features.CircleMarker(
                    [lat, lng],
                    radius=5,
                    color='red'
                    popup=label,
                    fill = True,
                    fill_color='red',
                    fill_opacity=0.6
                ).add_to(venues_map)
           folium.Circle([latitude, longitude], radius=250, fill=False, color='black').add_to(venues_map)
           folium.Circle([latitude, longitude], radius=500, fill=False, color='black').add_to(venues_map) folium.Circle([latitude, longitude], radius=1000, fill=False, color='black').add_to(venues_map)
           # display map
           venues_map
```

Out[360]: Make this Notebook Trusted to load map: File -> Trust Notebook

```
In [362]:
          venues_map = folium.Map(location=[latitude, longitude], zoom_start=15) # generate map centred around the met
          ro station
          # add a red circle marker to represent the metro station
          folium.features.CircleMarker(
              [latitude, longitude],
              radius=10,
              color='black',
              popup='Metro Station',
              fill = True,
              fill_color = 'black',
              fill_opacity = 0.6
          ).add_to(venues_map)
          for lat, lng, label in zip(dataframe_filtered.lat, dataframe_filtered.lng, dataframe_filtered.categories):
              folium.features.CircleMarker(
                  [lat, lng],
                  radius=5,
                  color='blue',
                  popup=label,
                  fill = True,
                  fill_color='blue',
                  fill_opacity=0.6
               ).add_to(venues_map)
          for lat, lng, label in zip(dataframe_filtered_bar.lat, dataframe_filtered_bar.lng, dataframe_filtered_bar.ca
              folium.features.CircleMarker(
                  [lat, lng],
                  radius=5,
                  color='red'
                  popup=label,
                  fill = True,
                  fill_color='red',
                  fill_opacity=0.6
              ).add_to(venues_map)
          for lat, lng, label in zip(dataframe_filtered_sport.lat, dataframe_filtered_sport.lng, dataframe_filtered_sp
          ort.categories):
              folium.features.CircleMarker(
                  [lat, lng],
                  radius=8,
                  color='green',
                  popup=label,
                  fill = True,
                  fill_color='green',
                  fill opacity=0.6
              ).add_to(venues_map)
          folium.Circle([latitude, longitude], radius=250, fill=False, color='black').add_to(venues_map)
          folium.Circle([latitude, longitude], radius=500, fill=False, color='black').add_to(venues_map)
          folium.Circle([latitude, longitude], radius=1000, fill=False, color='black').add_to(venues_map)
          # display map
          venues_map
```

## **Results and Discussion**

The analysis shows that although there is a good number of food places in the neighborhood (~33 in our area of interest which was 1km around the metro station), there are pockets of low food service density fairly close to sports centers. Highest concentration of restaurants was detected north and south from the metro station, so we focused our attention to east area, corresponding to a high concentration of sport centers and low food and drink services. Another borough was identified as potentially interesting (Peñagrande, south-west from the metro) as it has only one bar and two restaurants, but our attention was focused on Mirasierra which offers a combination of steady flow of people, closeness to the metro station, strong socio-economic dynamics and a low restaurant density.

Recommended zones should therefore be considered only as a starting point for more detailed analysis which could eventually result in location which has not only no nearby competition but also other factors taken into account and all other relevant conditions met.

# Conclusion

The purpose of this project was to identify areas close to Lacoma metro station with low number of restaurants (and bars) in order to starting a new food business exploiting the business opportunities due to Covid-19 impact through restaurants. There is the search for optimal location. By calculating restaurant density distribution from Foursquare data we have first identified general boroughs that justify further analysis (Mirasierra and Peñagrande) related to sport services around the areas.

Final decission on optimal restaurant location will be made by stakeholders based on specific characteristics of neighborhoods and locations in every recommended zone, taking into consideration additional factors like attractiveness of each location (proximity to park or water), levels of noise / proximity to major roads, real estate availability, prices, social and economic dynamics of every neighborhood etc.

Patricia Perez		
In [ ]:		