SOLVINGAN
OPTIMAL
BUSINESS
LOCATION
PROBLEM



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### Objective

In this project we will be determining the optimal location of a business in a city area by:

Extracting the necessary data from the Madrid's City Hall and the Foursquare API.

Making a segmentation by neighborhood and population characteristics in Madrid (Clustering).

Analyzing the results and extracting conclusions based on them.

The data that was used contained information about the nationalities of the inhabitants of each neighborhood and the quantity of people by neighborhood:

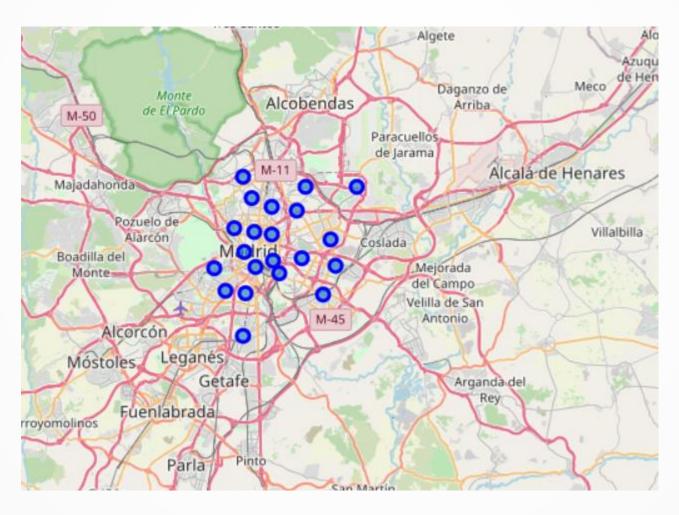
	Country of Procedence	Total Ciudad de Madrid	Centro	Arganzuela	Retiro	Salamanca	Chamartin	Tetuán	Chamberí	Fuencarral- El Pardo	Moncloa- Aravaca	Latina
0	Rumanía	45036.0	815.0	754.0	480.0	753.0	680.0	1468.0	597.0	1830.0	991.0	4904.0
1	China	37276.0	1508.0	1356.0	564.0	755.0	652.0	1988.0	816.0	1733.0	960.0	2554.0
2	Ecuador	23953.0	647.0	741.0	265.0	619.0	380.0	1395.0	453.0	632.0	387.0	2194.0
3	Venezuela	23359.0	1563.0	913.0	638.0	1564.0	933.0	1310.0	794.0	1428.0	630.0	1448.0
4	Colombia	22618.0	998.0	717.0	483.0	803.0	551.0	822.0	659.0	999.0	454.0	1786.0

# In order to access the Foursquare API data, the raw data had to be transformed this into something useful for the API:

- A second data frame was created
- The neighborhoods' names were included
- The latitude and longitude values of each neighborhood were added

	Neighborhood	Latitude	Longitude
0	Centro	40.415347	-3.707371
1	Arganzuela	40.402733	-3.695403
2	Retiro	40.408072	-3.676729
3	Salamanca	40.430000	-3.677778
4	Chamartin	40.453333	-3.677500
5	Tetuán	40.460556	-3.700000
6	Chamberí	40.432792	-3.697186
7	Fuencarral-El Pardo	40.478611	-3.709722
8	Moncloa-Aravaca	40.435151	-3.718765
9	Latina	40.402461	-3.741294

### Once the data was obtained, it was possible to draw the neighborhoods on a map:



## Then, the nearby venues by neighborhood were extracted along with their frequencies of occurrence

	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	9th Most Common Venue	10th Most Common Venue
0	Arganzuela	Restaurant	Spanish Restaurant	Grocery Store	Bakery	Tapas Restaurant	Gym / Fitness Center	Falafel Restaurant	Sandwich Place	Breakfast Spot	Burger Joint
1	Barajas	Hotel	Spanish Restaurant	Restaurant	Coffee Shop	Tapas Restaurant	Wine Bar	Diner	Mexican Restaurant	Fast Food Restaurant	Argentinian Restaurant
2	Carabanchel	Soccer Field	Bakery	Pizza Place	Plaza	Nightclub	Burger Joint	Fast Food Restaurant	Metro Station	Tapas Restaurant	Park
3	Centro	Plaza	Tapas Restaurant	Spanish Restaurant	Hostel	Cocktail Bar	Bistro	Café	Ice Cream Shop	Pastry Shop	Gym / Fitness Center
4	Chamartin	Spanish Restaurant	Restaurant	Grocery Store	Bakery	Tapas Restaurant	Café	Gastropub	Coffee Shop	Pizza Place	Park
5	Chamberí	Spanish Restaurant	Restaurant	Bar	Japanese Restaurant	Brewery	Café	Tapas Restaurant	Plaza	Italian Restaurant	Gastropub
6	Ciudad Lineal	Spanish Restaurant	Supermarket	Argentinian Restaurant	Gastropub	Restaurant	Burger Joint	Gym / Fitness Center	Cosmetics Shop	Music Venue	Diner
7	Fuencarral-El Pardo	Clothing Store	Fast Food Restaurant	Burger Joint	Italian Restaurant	Tapas Restaurant	American Restaurant	Coffee Shop	Sandwich Place	Restaurant	Pizza Place

Finally, a data frame containing the most common venues by neighborhood was created:



#### Conclusions

- As far as we can see with this data, there are no Mexican populations registered in Madrid. However, in Cluster 1, it is possible to notice that there's a Mexican restaurant located in the "Centro" neighborhood, which is the town center.
- If a deeper exam is performed into this cluster, it is noticeable that its living population are mostly Latinos, mixed with some other Europeans, but mainly, the people living in this cluster come from south American countries. Apart of this fact, other kinds of Latin restaurants can be found, like Argentinian restaurants, tapas restaurants, and Italian restaurants. So, it is possible to tell that the inhabitants of this area like these kinds of food.
- By following this logic, if we would like to open a new Mexican restaurant in the city or any kind of restaurant in fact, it would only be necessary to find a where are the restaurants similar the one we want to open, study the population in that area, and find similar clusters of population in the city that don't have yet or have very few restaurants like the one we would like to open.

#### Conclusions - Cont'd

- In this example, clusters 4 and 5 could make a good match for our target population. Looking at the venues in these clusters, it is possible to find one Mexican restaurant, and a good bunch of fast food, Argentinian, and south American restaurants. So, in these clusters, it is possible to state that the existing restaurants matches the population's nationalities and tastes.
- In conclusion and taking into consideration the explanations given above as well as the data, it is highly possible that clusters 4 and five could be a good place to open our Mexican restaurants. As explained above, the same logic could apply to open other kind of restaurant or business in any other area of the city. It is only necessary to examine the existing businesses in our target area, and study the population, then compare these two factors with the same ones in areas where there are existing businesses like the one we want to open, and then verify if the matching is correct.

Thank you.