

ANALYSIS OF RESIDENTIAL NEIGHBORHOODS IN MADRID, SPAIN

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DISCOVERING RESIDENTIAL NEIGHBORHOODS IS VALUABLE FOR FAMILIES

- ALL CITIES (ESPECIALLY THE BIGGEST ONES) HAVE DIFFERENT TYPE OF NEIGHBORHOODS. SOME OF THEM ARE MORE SUITABLE FOR TOURIST, OTHER ONES FOR SHOPPING AND SOME OF THEM FOR SETTING A FAMILY.
- FAMILIES NEED TO KNOW WHICH ARE THE MOST SUITABLE NEIGHBORHOODS FOR LIVING IN A NEW CITY
- MADRID IS THE CAPITAL CITY OF SPAIN. IT HAS MORE THAN 100 NEIGHBORHOODS WITH A TOTAL POPULATION OF 3.334.730. IT IS PRETTY DIFFICULT FOR A NEW FAMILY THAT ARRIVES TO THE CITY TO KNOW WHERE TO START LIVING.
- THE OBJECTIVE OF THIS PROJECT IS TO DIFFERENTIATE WHICH ARE THE RESIDENTIAL NEIGHBORHOODS IN THIS CITY.

DATA ACQUISITION AND CLEANING

- INFORMATION OF EVERY NEIGHBORHOOD IN MADRID, SUCH AS THE NAME AND THE PER INCOME CAPITA FROM:
[HTTPS://WWW.MADRID.ES/PORTALES/MUNIMADRID/ES/INICIO/EL-AYUNTAMIENTO/ESTADISTICA/?VGNEXTFMT=DEFAULT&VGNEXTCHANNEL=8156E39873674210VGNVCM1000000B205A0ARCRD](https://www.madrid.es/portales/munimadrid/es/Inicio/El-Ayuntamiento/Estadistica/?vgnextfmt=default&vgnextchannel=8156E39873674210VGNVCM1000000B205A0ARCRD).
- GEOGRAPHICAL COORDINATES OF EVERY NEIGHBORHOOD USING THE GEOPY LIBRARY.
- INFORMATION OF PUBLIC SCHOOLS IN THE CITY FROM : [HTTPS://DATOS.MADRID.ES/PORTAL/SITE/EGOB/](https://datos.madrid.es/portal/site/egob/)

METHODOLOGY

- COLLECT THE MADRID CITY DATA IN DATAFRAMES:

	Neighborhood	Renta	Latitude	Longitude	Colegios
0	Abrantes	10544.0	40.37980	-3.72636	4
1	Acacias	19323.0	40.40137	-3.70669	1
2	Adelfas	18991.0	40.40173	-3.67288	1
3	Aeropuerto	9814.0	40.48337	-3.55949	0
4	Alameda de Osuna	19871.0	40.45818	-3.58953	1
...
126	Ventas	12072.0	40.42238	-3.65020	3
127	Villaverde Alto, C.H. Villaverde	9354.0	40.34922	-3.71211	0
128	Vinateros	12695.0	40.40444	-3.64029	2
129	Vista Alegre	10775.0	40.38492	-3.74635	2
130	Zofio	9601.0	40.37987	-3.71495	2

131 rows x 5 columns

- USING FOUR SQUARE API WE WILL FIND ALL VENUES FOR EACH NEIGHBORHOOD.
- CLUSTER THE NEIGHBORHOODS ACCORDING TO ITS VENUES.
- VISUALIZE THE CLUSTER AND COMPARING TO PUBLIC SCHOOLS DISTRIBUTION

METHODOLOGY

- COLLECT THE MADRID CITY DATA IN DATAFRAMES:
- USING FOUR SQUARE API WE WILL FIND ALL VENUES FOR EACH NEIGHBORHOOD:

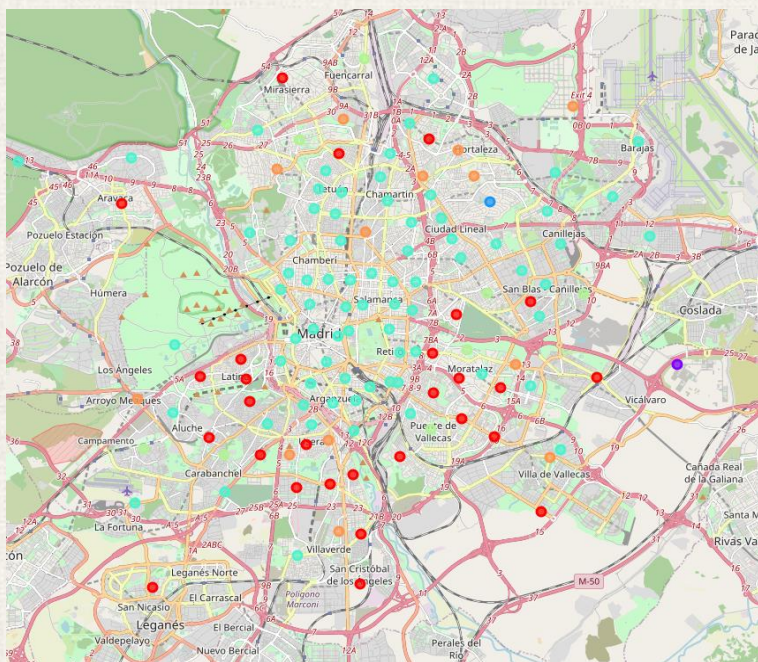
```
#create list with neighborhoods to exclude
neigh_to_exclude = Madrid_venues_count[Madrid_venues_count['Venue Category Count'] < 4]
#create filtered dataframe by excluding neighborhoods in above list
Madrid_venues_filt = Madrid_venues[~Madrid_venues['Neighborhood'].isin(neigh_to_exclude)]
#rename filtered dataframe back to toronto_venues
Madrid_venues = Madrid_venues_filt
#check counts after filtering
Madrid_venues.groupby('Neighborhood').count()
```

Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude
Abrantes	7	7	7	7	7
Acacias	54	54	54	54	54
Adelfas	50	50	50	50	50
Alameda de Osuna	23	23	23	23	23
Almagro	100	100	100	100	100
...
Ventas	11	11	11	11	11
Villaverde Alto, C.H. Villaverde	4	4	4	4	4
Vinateros	8	8	8	8	8
Vista Alegre	17	17	17	17	17
Zofio	7	7	7	7	7

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METHODOLOGY

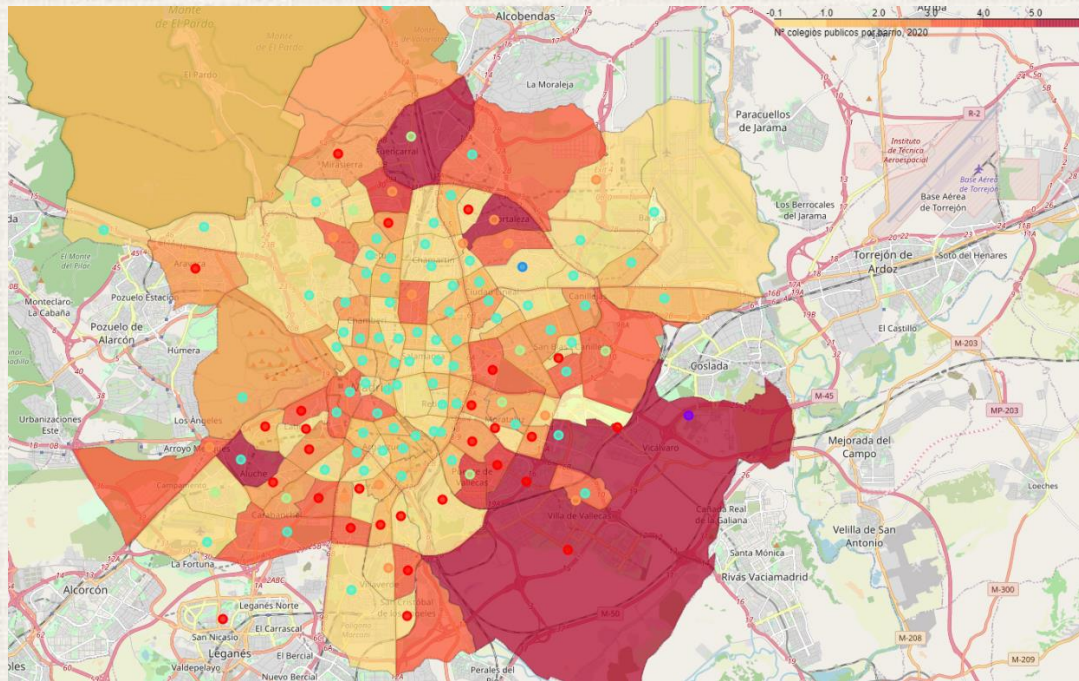
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- **CLUSTER THE NEIGHBORHOODS ACCORDING TO ITS VENUES:**



- VISUALIZE THE CLUSTER AND COMPARING TO PUBLIC SCHOOLS DISTRIBUTION

METHODOLOGY

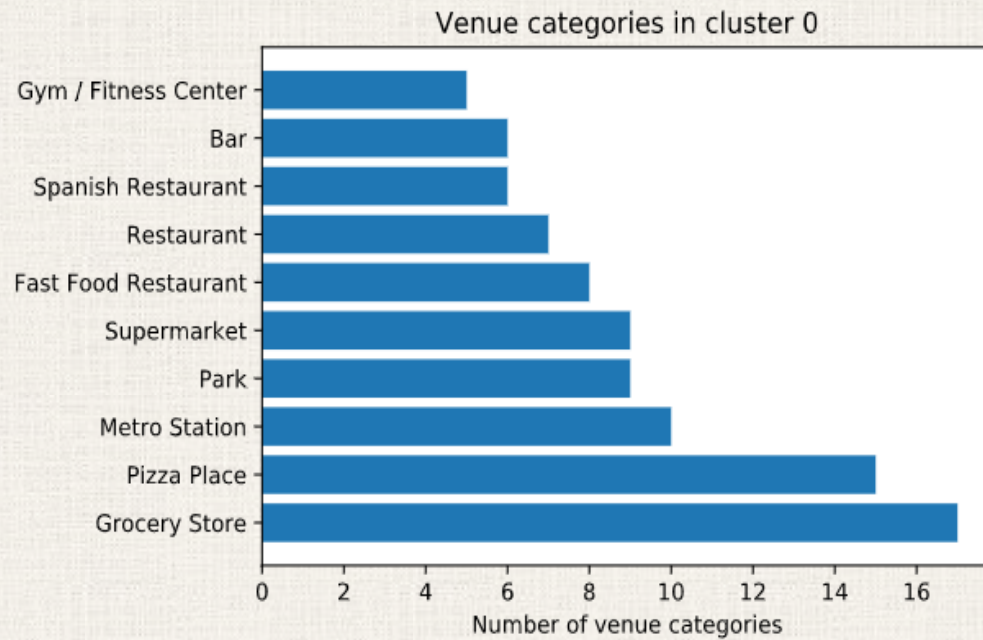
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- **VISUALIZE THE CLUSTER AND COMPARING TO PUBLIC SCHOOLS DISTRIBUTION:**



RESULTS

- MOST SUITABLE RESIDENTIAL NEIGHBORHOODS:

Cluster 0: Residential



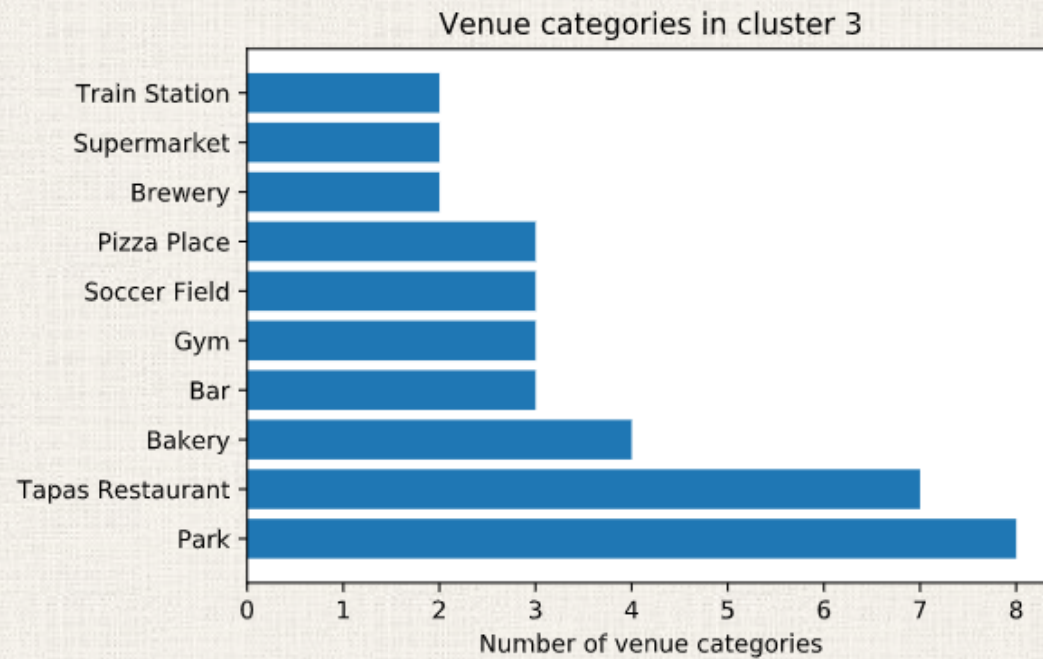
Plenty of useful services for a family like:

- *Grocery stores*
- *Metro stations*
- *Parks*
- *Supermarkets*
- *Gyms*

RESULTS

- MOST SUITABLE RESIDENTIAL NEIGHBORHOODS:

Cluster 3: Sport lovers residential area



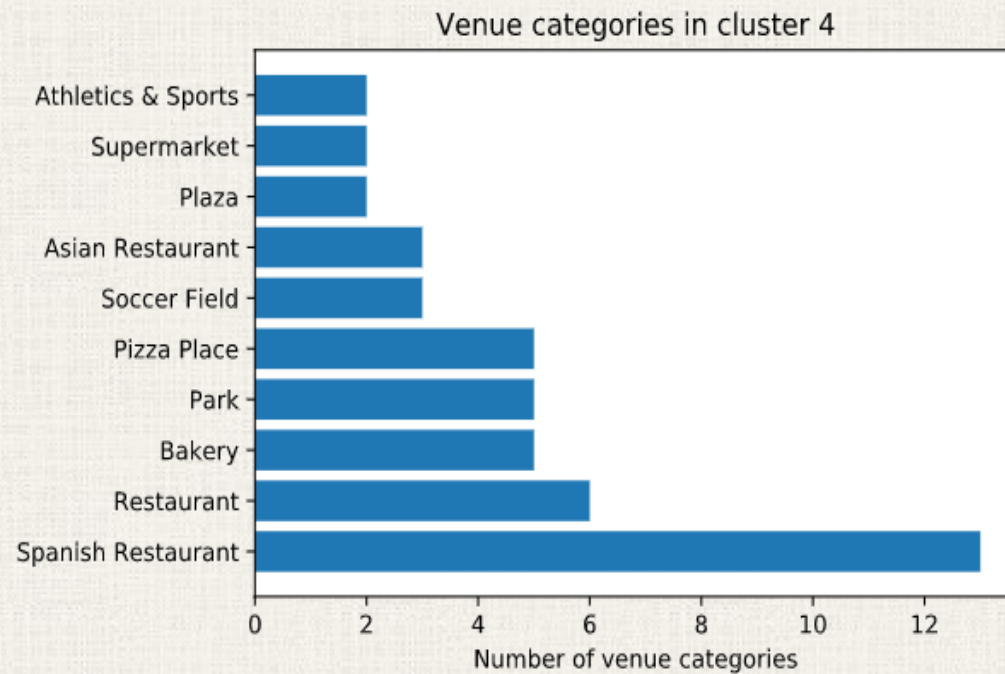
Interesting services for a sport lover family like:

- *Soccer fields*
- *Gyms*
- *Parks*
- *Supermarkets*
- *Train stations*

RESULTS

- MOST SUITABLE RESIDENTIAL NEIGHBORHOODS:

Cluster 4: Food lovers residential area



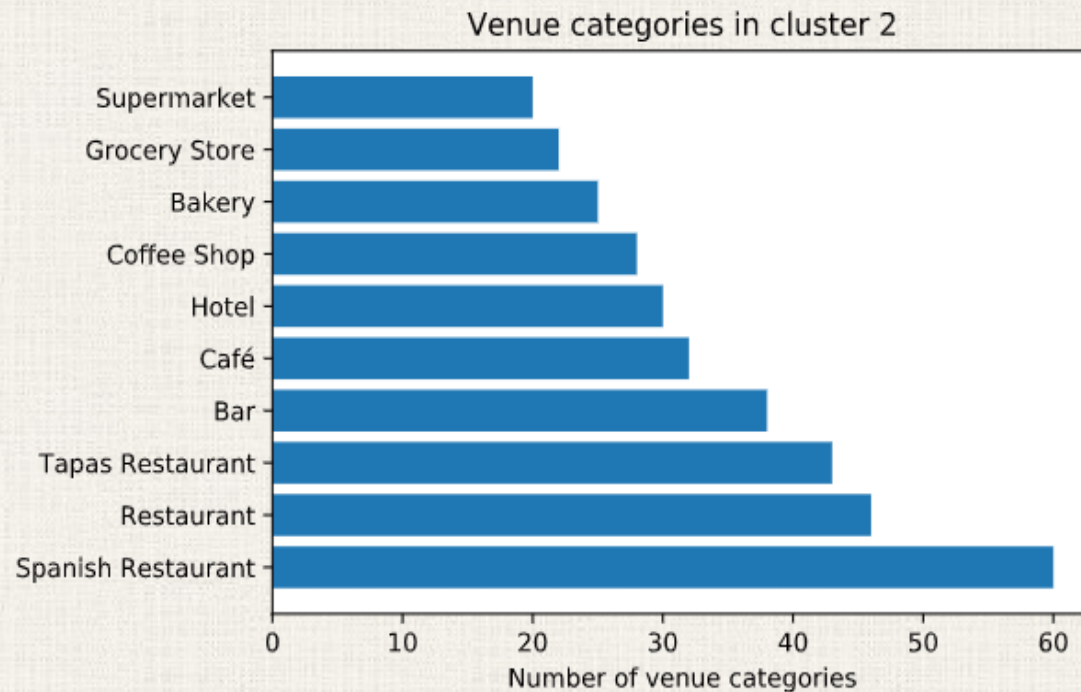
Plenty of useful services for a family like:

- *Restaurants*
- *Spanish restaurants*
- *Parks*
- *Supermarkets*
- *Plaza*

RESULTS

- LESS SUITABLE RESIDENTIAL NEIGHBORHOODS:

● *Cluster 2: Downtown*

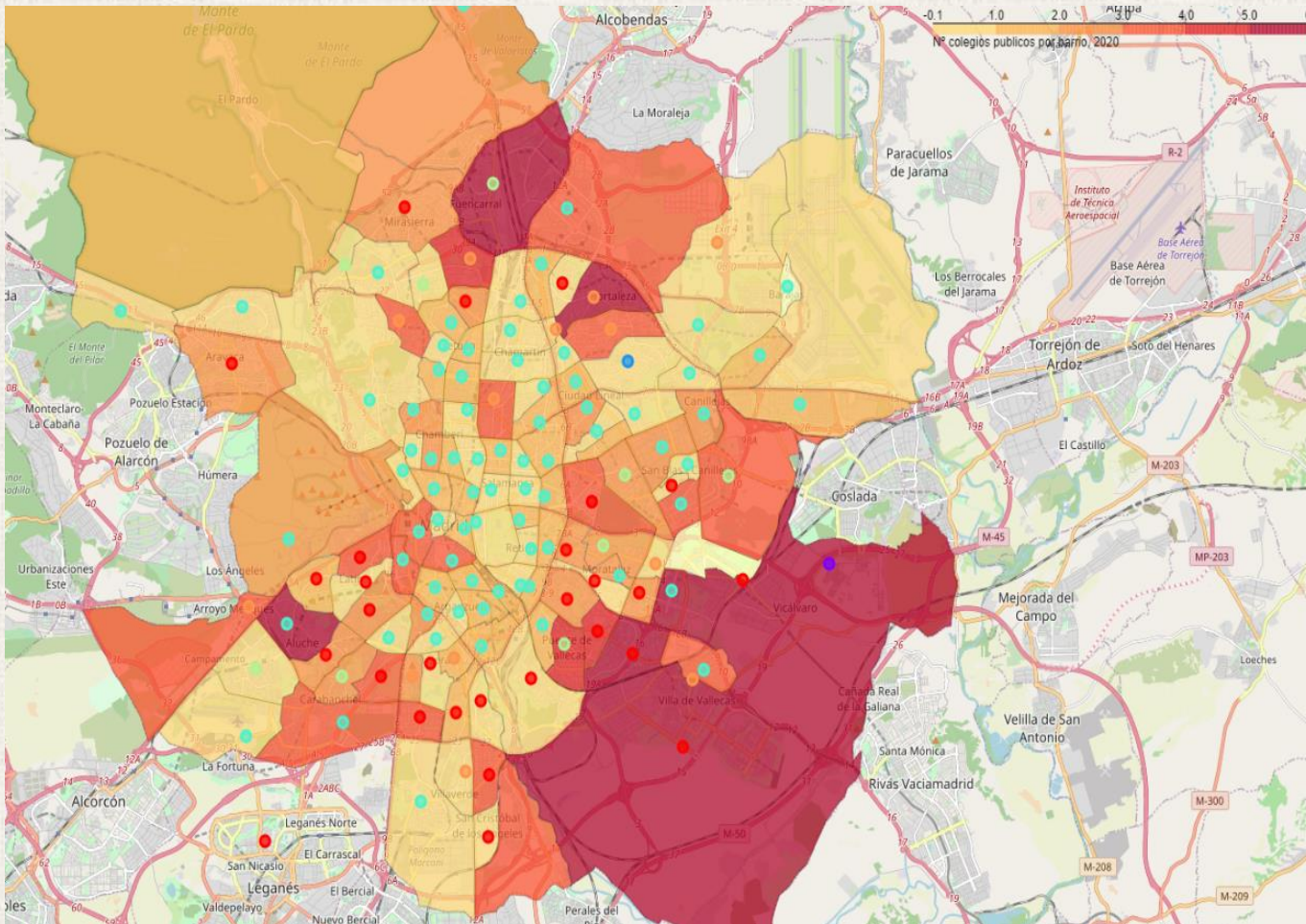


Plenty of useful services for a tourist like:

- *High end Restaurants*
- *Hotels*
- *Coffee shops*
- *Bar*

CONCLUSION

- COMPARING TO PUBLIC SCHOOLS DISTRIBUTION:



We can see that the previous clusters (●, ● and ●) are in those neighborhoods with a huge amount of public schools.

That is according to what is expected so it seems to be a pretty reliable Project.

As we have shown before, we can even decide among different type of residential neighborhoods.