

The background of the entire slide is a dense, overlapping field of three-dimensional numbers. The numbers are rendered in a light blue color with a soft gradient and are cast into deep shadows, giving them a strong sense of depth and volume. They are scattered across the frame in various orientations, some standing upright and others tilted, creating a complex, textured visual effect. The numbers include digits from 0 to 9, with some appearing more frequently than others.

The Battle of Neighborhoods - Chile

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Introduction / Business Problem

- ◆ In this project we will try to find an optimal location for a Hotel. Specifically, this report will be targeted to stakeholders interested in opening a new Hotel in one of the regional capital cities of Chile.
- ◆ Since there are not lots of hotels in Chile, we will try to detect locations that are not already crowded with hotels. We are also particularly interested in areas with no Hotels in the principal venues.
- ◆ We will use our data science powers to generate a few most promising cities based on these criteria. Advantages of each area will then be clearly expressed so that best possible final location can be chosen by stakeholders.

Data

- ◆ Based on definition of our problem, factors that will influence our decision are:
 - ◆ Location of each regional capital city in Chile
 - ◆ List of the venues of each capital city.
- ◆ The data used for this project was taken from the Wikipedia web page https://es.wikipedia.org/wiki/Anexo:Ciudades_de_Chile, which was processed in Excel to obtain the table with the latitude and longitude for each capital city.
- ◆ The Foursquare API search feature will be used to collect the venues registered in each capital city with and specific radius from the location.
- ◆ In addition, various Python packages will be used to create maps and machine learning models to understand the data collected and give the best advice for the business problem.

Methodology

1. Get the data from the Wikipedia Web Page using excel to export the table with the location from the capital cities of each region in Chile.
2. Export the data to the Jupiter Notebook in IBM Watson Studio
3. Using the location (latitude and longitude) of each regional capital in Chile and the Foursquare API feature, get the venues for a radius of 40 km around of the city location.
4. Folium visualization library will be used to show the location of the capital cities in Chile.
5. K-means machine learning will be used to create 5 cluster depending of the venues of each capital city.
6. The clusters will be showed at the same map created before for the capital cities in Chile, but at this time in different cluster.
7. Analise each cluster and look for the best to open a new hotel in Chile.

Results

- Even tough, the radius used for the analysis was 40 km and the top of venues was 100, the number of venues obtained for each capital city was 100 just for 6 of the 13 cities in Chile. This means that the Foursquare data for this country is not sufficient to complete the requirements.
- Even so, I continued with the analysis to check which are the most common venues used in Chile and the results were the following.

| | City_fixed | 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 | Antofagasta | Beach | Restaurant | Chinese Restaurant | Bistro | Park | Peruvian Restaurant | Seafood Restaurant | Hotel | Soccer Field | Pizza Place |
| 1 | Arica | Restaurant | Beach | Hotel | Ice Cream Shop | Surf Spot | Chinese Restaurant | History Museum | South American Restaurant | Pub | Vegetarian / Vegan Restaurant |
| 2 | Calama | Hotel | Restaurant | Breakfast Spot | Gym | Fast Food Restaurant | Historic Site | Convenience Store | Nightclub | Chinese Restaurant | Mountain |
| 3 | Chillan | Plaza | Restaurant | Sushi Restaurant | Pizza Place | Coffee Shop | Park | Hotel | Peruvian Restaurant | Café | Asian Restaurant |
| 4 | Concepcion | Beach | Sandwich Place | Pizza Place | Italian Restaurant | Hotel | Park | City | Ice Cream Shop | Sushi Restaurant | Plaza |
| 5 | Copiapó | Hotel | Nightclub | Pizza Place | Café | History Museum | Diner | Restaurant | Pub | Sushi Restaurant | Movie Theater |
| 6 | Coyhaique | Bed & Breakfast | Café | Restaurant | Park | Pub | Sushi Restaurant | BBQ Joint | Scenic Lookout | Sandwich Place | Hotel |
| 7 | Iquique | Hotel | Plaza | Ice Cream Shop | Beach | Latin American Restaurant | Restaurant | Museum | Pizza Place | Sandwich Place | Coffee Shop |
| 8 | La Serena | Seafood Restaurant | Restaurant | Beach | Burger Joint | Surf Spot | Pizza Place | Dessert Shop | Spa | Sandwich Place | Pool |
| 9 | Puerto Montt | Hotel | Restaurant | Café | Scenic Lookout | Beach | Boat or Ferry | German Restaurant | History Museum | Rental Car Location | Bed & Breakfast |
| 10 | Punta Arenas | Restaurant | Hotel | Café | History Museum | Other Great Outdoors | Hostel | Scenic Lookout | Tea Room | Gastropub | Diner |
| 11 | Rancagua | Restaurant | Pizza Place | Hotel | BBQ Joint | Soccer Field | Plaza | Gastropub | Gas Station | Ice Cream Shop | Sandwich Place |
| 12 | Santiago | Park | Bakery | Pizza Place | Vineyard | Golf Course | Snack Place | Scenic Lookout | Deli / Bodega | Museum | Mountain |
| 13 | Talca | Park | Coffee Shop | Ice Cream Shop | Vineyard | Pizza Place | Other Great Outdoors | Cocktail Bar | Resort | Restaurant | Bookstore |
| 14 | Temuco | Plaza | Café | Sandwich Place | Sushi Restaurant | Tea Room | Pizza Place | Diner | Burger Joint | Restaurant | Coffee Shop |
| 15 | Valdivia | Restaurant | Scenic Lookout | Hotel | Bed & Breakfast | Beach | Bar | Café | Brewery | Historic Site | Park |



Results

- ◇ Looking at the results for the 10 most common venues in the capital cities of Chile, it is possible to see into the 3rd first categories the venue Hotel which means that although the data created for Chile is not enough, Hotels are one of the most popular to include in Foursquare.
- ◇ After this, I used k-means machine learning to divide the cities in 5 different clusters depending of the venues of each one.

Conclusion

- ◇ After clustering the regional capital cities of Chile was possible to realise that Calama and Copiapo are the cities where Hotel is the most popular venues. It could be possible because the main industry in both cities is the mining industry. This could be produced because a lot of people use the flight in and flight out system, demanding a lot of accommodation places.
- ◇ In addition, the data obtained from the Foursquare API is not good enough to do a deep analysis for the venues in the capital cities in Chile. Just 6 of the 13 cities got the maximum number of venues required which means that the data for the rest of the cities is not enough or that not too many people use this app in the country.

| | City | 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 |
|---|---------|----------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|------------------------|
| 3 | Calama | 3 | Hotel | Restaurant | Breakfast Spot | Gym | Fast Food Restaurant | Historic Site | Convenience Store | Nightclub | Chinese Restaurant | Mountain |
| 4 | Copiapo | 3 | Hotel | Nightclub | Pizza Place | Café | History Museum | Diner | Restaurant | Pub | Sushi Restaurant | Movie Theater |