Clustering in North San Antonio and Surrounding Area

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1. Introduction

1.1 Background

San Antonio, TX is a diverse city that is seeing continued growth, and the surrounding areas are also experiencing an increase in development. It was recently ranked as one of the fastest growing cities in the United States. The intent of this project is to identify areas in north San Antonio, TX and the surrounding area that would be ideal for opening a restaurant. We will explore the current offerings and review the potential for future business endeavors. This analysis would be particularly helpful for a restaurateur who is interested in starting or expanding their business.

1.2 Business Problem

The objective is to find the ideal location for a restaurateur to open a new restaurant in north San Antonio or the surrounding area.

1.2 Interest

Developers, Franchise owners, entrepreneurs, and other restaurateurs would be interested in understanding the make-up of San Antonio and the surrounding areas.

2. Data Acquisition and Cleaning

2.1 Data Used and the Sources

We will identify the major neighborhoods/ sub areas to explore after reviewing the San Antonio map (produced with Folium) and then create a dataframe that includes the neighborhood name, along with the latitude and longitude that are obtained with the Geocoder package.

We will be exploring the venue category by neighborhood. We will use the Foursquare API to obtain venue data related to these neighborhoods/ sub-areas.

3. Methodology

3.1 Visualizing the neighborhoods to explore

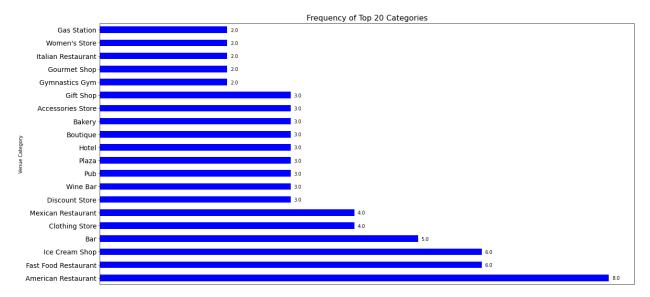
The Folium package was first used to review the smaller cities within the north San Antonio area, and it was used again after selecting the top cities to review to visualize where they are located.



3.2 Reviewing nearby venues

Foursquare API data was used to access the data of venues within a 500m radius. The below graph is a subset of the data that shows the frequency of the top 20 venue categories in the area.

This is helpful for understanding what kind of areas we're looking at and what restaurants might be most successful (or could already be saturated).



The top 10 most common venues by neighborhood were also reviewed.

10th Most Common Venue	9th Most Common Venue	8th Most Common Venue	7th Most Common Venue	6th Most Common Venue	5th Most Common Venue	4th Most Common Venue	3rd Most Common Venue	2nd Most Common Venue	1st Most Common Venue	Neighborhood	
Gourmet Shop	Snack Place	Mexican Restaurant	Hotel	Bistro	Restaurant	Burger Joint	Café	American Restaurant	Boutique	Boerne, TX	0
Cosmetics Shop	Convenience Store	Construction & Landscaping	Coffee Shop	Cocktail Bar	Discount Store	American Restaurant	Airport	Gift Shop	Wine Bar	Bulverde, TX	1
Gymnastics Gym	Ice Cream Shop	Dive Bar	Italian Restaurant	Locksmith	Convenience Store	Construction & Landscaping	Business Service	BBQ Joint	Gas Station	Converse, TX	2
American Restaurant	Video Store	Gas Station	Bar	Burger Joint	Ice Cream Shop	Fast Food Restaurant	Smoke Shop	Pharmacy	Discount Store	Helotes, TX	3
Mobile Phone Shop	Mexican Restaurant	Shoe Store	Chinese Restaurant	Pharmacy	Other Repair Shop	Automotive Shop	Pawn Shop	Ice Cream Shop	Fast Food Restaurant	Leon Valley, TX	4
Clothing Store	Gourmet Shop	Cocktail Bar	Plaza	Hotel	Market	American Restaurant	German Restaurant	Pub	Bar	New Braunfels, TX	5
Convenience Store	Construction & Landscaping	Coffee Shop	Cocktail Bar	Clothing Store	Bakery	Gymnastics Gym	Weight Loss Center	Baseball Field	Pizza Place	Schertz, TX	6
Juice Bar	Massage Studio	Gastropub	Fast Food Restaurant	Cosmetics Shop	Sandwich Place	American Restaurant	Women's Store	Clothing Store	Accessories Store	Stone Oak, TX	7
Construction & Landscaping	Coffee Shop	Cocktail Bar	Clothing Store	Chinese Restaurant	Food Truck	Frozen Yogurt Shop	Plaza	Golf Course	Discount Store	Windcrest, TX	8

3.3 Clustering

K-means clustering, which is an unsupervised machine learning technique, was used to divide the areas into three clusters. The dataframe with the restaurant venues was transformed using one-hot encoding. The process of one-hot is used for converting categorical variables into a form that can be used in a machine learning algorithm. Including a sample of the resulting df and supporting code.



4. Results

Three clusters were created. As shown in the map below, one cluster is in red, one is in blue, and one is in purple.



Cluster 1 contains the most areas, and the neighborhoods in this group appear to have more restaurants in their top 10 common venues.

Cluster 2 contains fewer restaurants, but it contains dessert and drink locations too.

Cluster 3 appears to be somewhat similar to Cluster 3, in that it does not have many restaurants. However, it contains an American restaurant.

5. Discussion

Based on the results, entrepreneurs and restaurateurs have several options, depending on their area of expertise. If they are not set on one particular cuisine though, my recommendation would be to open an American food restaurant in cluster 2 (New Braunfels area).

We saw from the initial exploratory analysis of all of the San Antonio and surrounding area results that American food restaurants are the most popular and fall into the top-rated categories. Cluster 1 has many restaurants already, and this might indicate that it could already be saturated with restaurants. Cluster 3 has fewer restaurants, but it does contain an American restaurant in its top venues. This leaves Cluster 2 (New Braunfels area) as the most desirable place to open an American food restaurant.

6. Conclusion

The San Antonio area continues to grow and develop. While outside the scope of this project, further analysis could include expanding the geographic constraints to include the entirety of San Antonio and surrounding areas, as opposed to just the north side. This could potentially help with increasing the size of the clusters. Using the tools learned in this course, we were able to create clusters and provide the recommendation for a restaurateur to open an American cuisine restaurant in the New Braunfels area.