



# **Clustering the different Neighborhoods in Metro Manila Using K-Means Clustering Algorithm**

# Businesses

## Aim & Advantages

# People

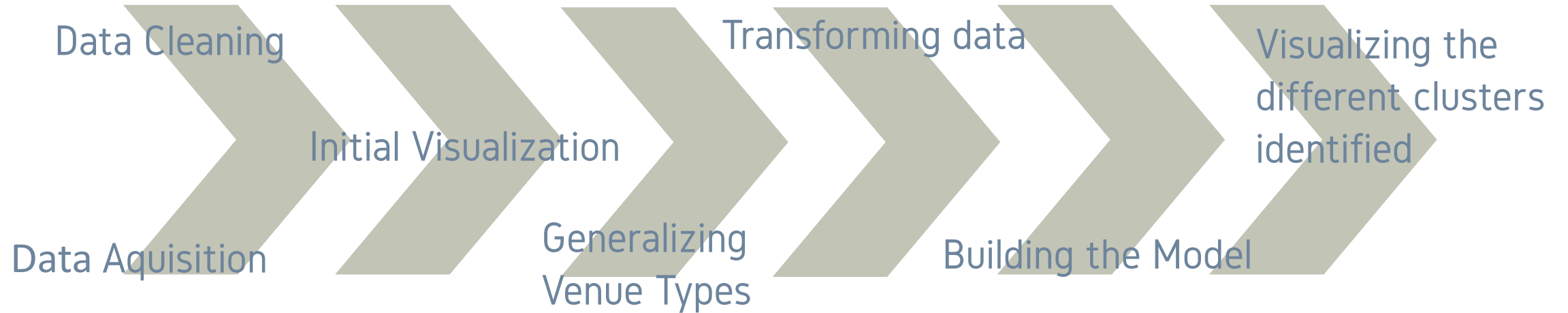
Customer targeting

Finding areas for new business  
locations

Knowledge to identify  
what neighborhoods match  
the lifestyle they desire.

# Methodology

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# Data Acquisition and Cleaning

List of  
Neighborhoods in  
Metro Manila

Details and Types of  
Venues in Metro  
Manila

Geospatial Data of  
the Neighborhoods in  
Metro Manila

Wikipedia

FourSquare API

Nominatim API

# Data Stored in Pandas Data Frame

Data frame with geospatial data of all the Neighborhoods

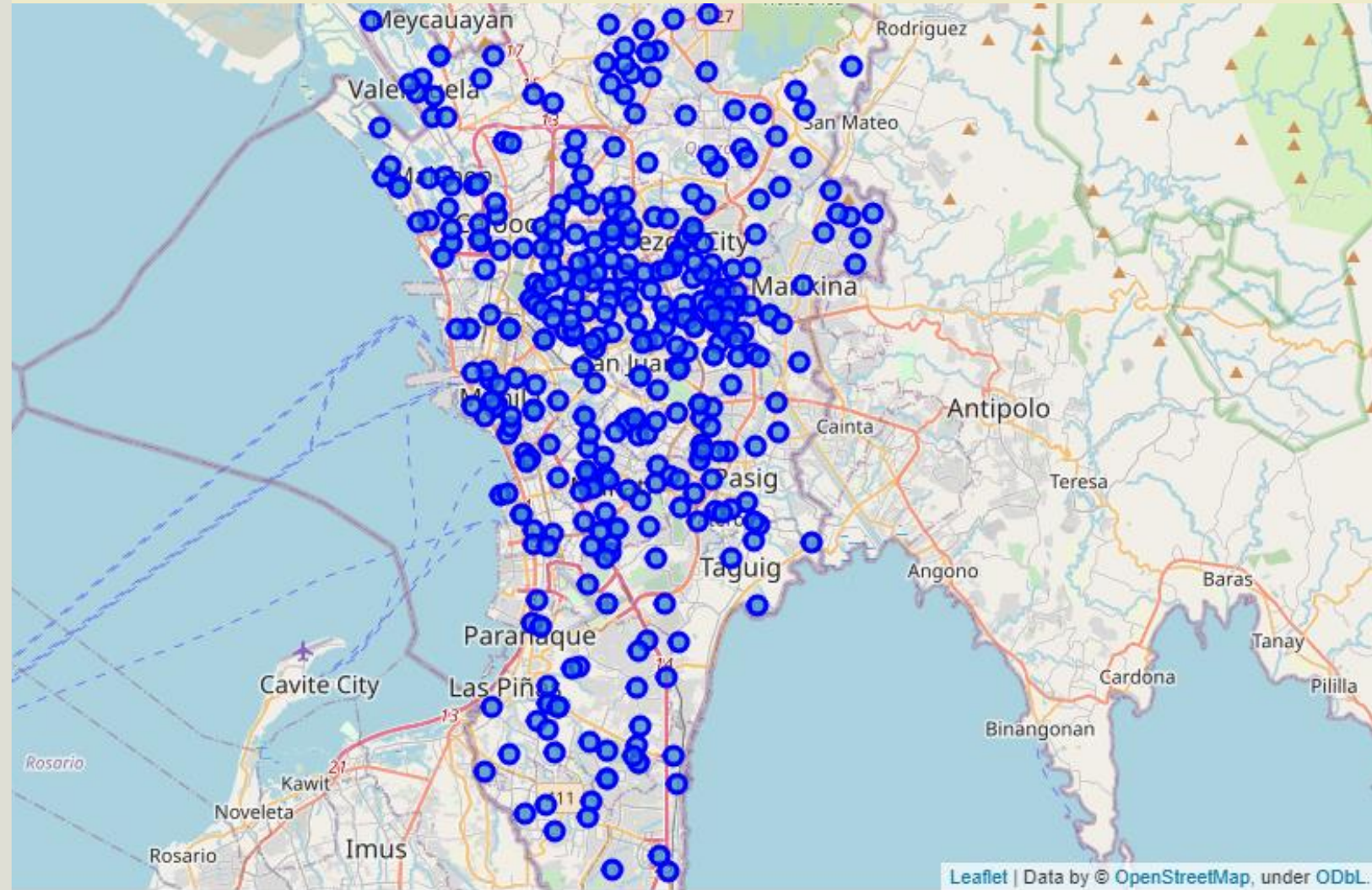
|   | ZIP code | Neighborhood            | City       | Latitude  | Longitude  |
|---|----------|-------------------------|------------|-----------|------------|
| 0 | 401      | Asian Development Bank  | San Juan   | 14.588076 | 121.058301 |
| 1 | 550      | Febias College of Bible | Valenzuela | 14.687899 | 120.981408 |
| 2 | 702      | Citibank                | Makati     | 14.607314 | 121.078924 |
| 3 | 704      | Producers Bank          | Makati     | 14.529699 | 121.041247 |
| 4 | 707      | Canadian Embassy        | Makati     | 14.560645 | 121.016578 |

Data frame details of the venues found in all the neighborhoods

|   | Neighborhood           | Neighborhood Latitude | Neighborhood Longitude | Venue                       | Venue Latitude | Venue Longitude | Venue Category           |
|---|------------------------|-----------------------|------------------------|-----------------------------|----------------|-----------------|--------------------------|
| 0 | Asian Development Bank | 14.588076             | 121.058301             | Craft Coffee Revolution     | 14.585859      | 121.059212      | Coffee Shop              |
| 1 | Asian Development Bank | 14.588076             | 121.058301             | The Nostalgia Dining Lounge | 14.587519      | 121.059753      | Restaurant               |
| 2 | Asian Development Bank | 14.588076             | 121.058301             | Gino's Brick Oven Pizza     | 14.585791      | 121.059640      | Pizza Place              |
| 3 | Asian Development Bank | 14.588076             | 121.058301             | The Café Mediterranean      | 14.585967      | 121.057122      | Mediterranean Restaurant |
| 4 | Asian Development Bank | 14.588076             | 121.058301             | Wildflour Café + Bakery     | 14.585866      | 121.059573      | Café                     |



# Initial Mapping of the Neighborhoods in Metro Manila



# Generalizing the specific Venue Types

337 different venues types  
from the FourSquare Dataset

Consolidating similar venue  
types

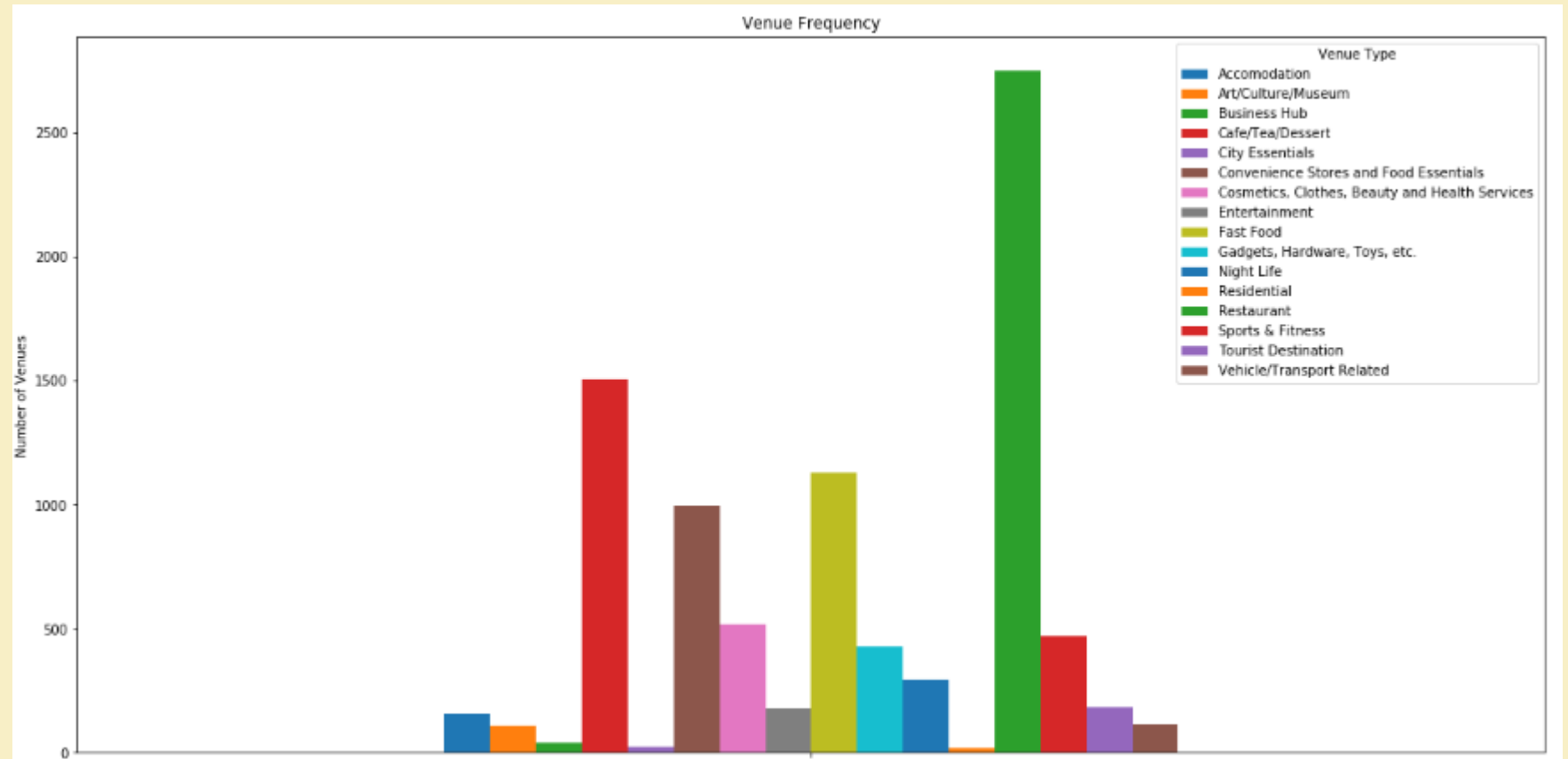
| Venue Type    | Venue Category       |
|---------------|----------------------|
| Residential   | Boarding House       |
|               | Residential Building |
|               | (Apartment/Condo)    |
| Venue Type    | Venue Category       |
| Accommodation | Hostel               |
|               | Hotel                |
|               | Hotel Pool           |
|               | Lounge               |
|               | Motel                |
|               | Rest Area            |
|               | Roof Deck            |
|               | Travel Lounge        |

| General Venue Types |  |
|---------------------|--|
| 1                   | Accommodation                                  |
| 2                   | Art/Culture/Museum                             |
| 3                   | Business Hub                                   |
| 4                   | Café/Tea/Dessert                               |
| 5                   | City Essentials                                |
| 6                   | Convenience stores and Food Essentials         |
| 7                   | Cosmetics, Clothes, Beauty and Health Services |
| 8                   | Entertainment                                  |
| 9                   | Fast Food                                      |
| 10                  | Gadgets, Hardware, Toys, etc.                  |
| 11                  | Night Life                                     |
| 12                  | Residential                                    |
| 13                  | Restaurants                                    |
| 14                  | Sports & Fitness                               |
| 15                  | Tourist Destination                            |
| 16                  | Vehicle/Transport Related                      |

# Visualization of the Different Venue Types

Show a large bias towards restaurants which is expected of a Metropolitan Region

Skews the model





# Applies an Importance Factor

Emphasizes lower count venue types

Shows a large prominence on neighborhoods that have high density of the low count venue types

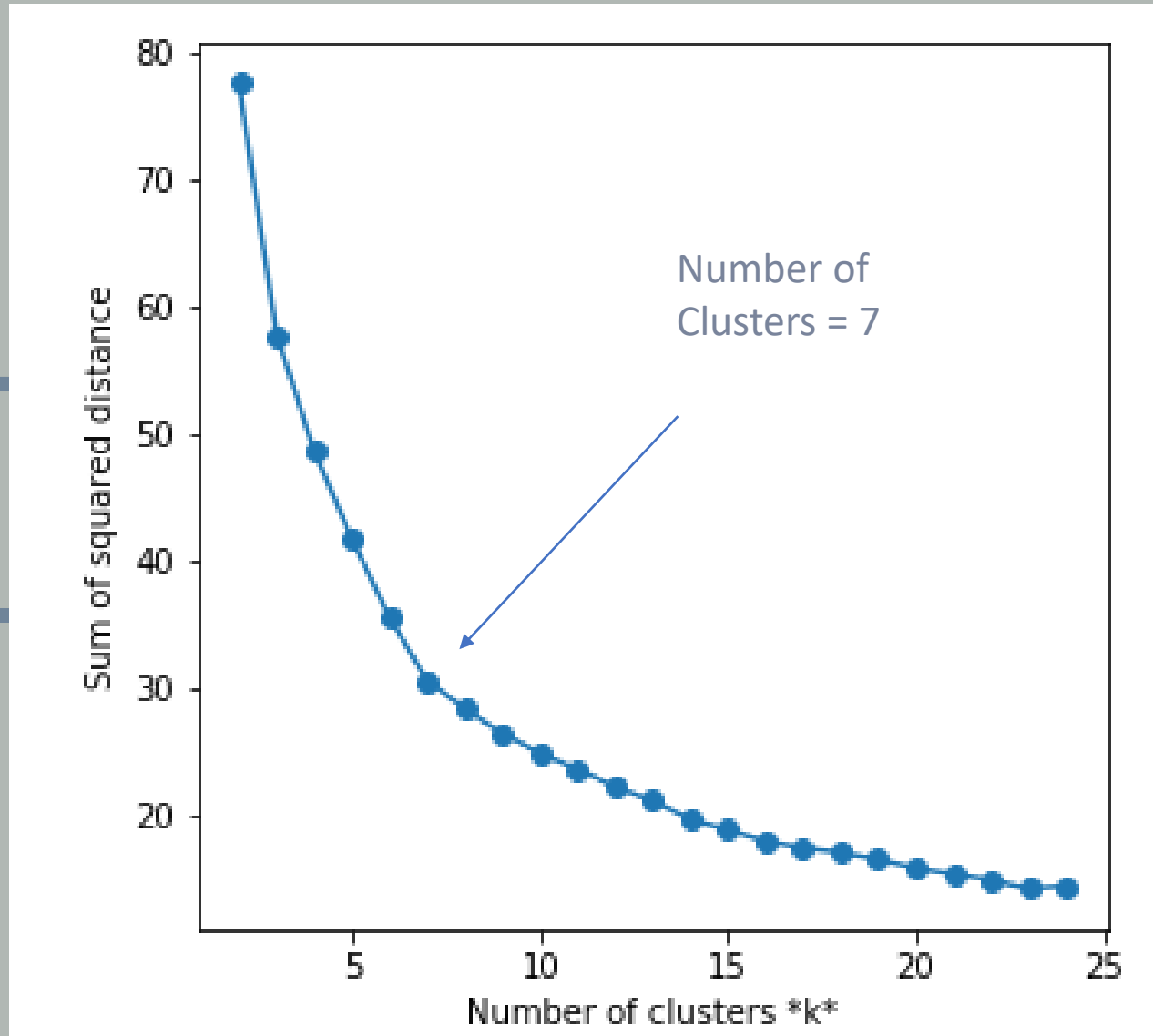
Deemphasize high venue count types

| Importance factor                              |          |
|--|----------|
| Venue Type                                     |          |
| Accomodation                                   | 0.107595 |
| Art/Culture/Museum                             | 0.155963 |
| Business Hub                                   | 0.414634 |
| Cafe/Tea/Dessert                               | 0.011296 |
| City Essentials                                | 0.653846 |
| Convenience Stores and Food Essentials         | 0.017068 |
| Cosmetics, Clothes, Beauty and Health Services | 0.032692 |
| Entertainment                                  | 0.096591 |
| Fast Food                                      | 0.015084 |
| Gadgets, Hardware, Toys, etc.                  | 0.040000 |
| Night Life                                     | 0.057627 |
| Residential                                    | 1.000000 |
| Restaurant                                     | 0.006182 |
| Sports & Fitness                               | 0.036247 |
| Tourist Destination                            | 0.093923 |
| Vehicle/Transport Related                      | 0.150442 |

# Transform Data using One-hot Encoding

# Finding the best K for K-Means Clustering

Using Elbow Method

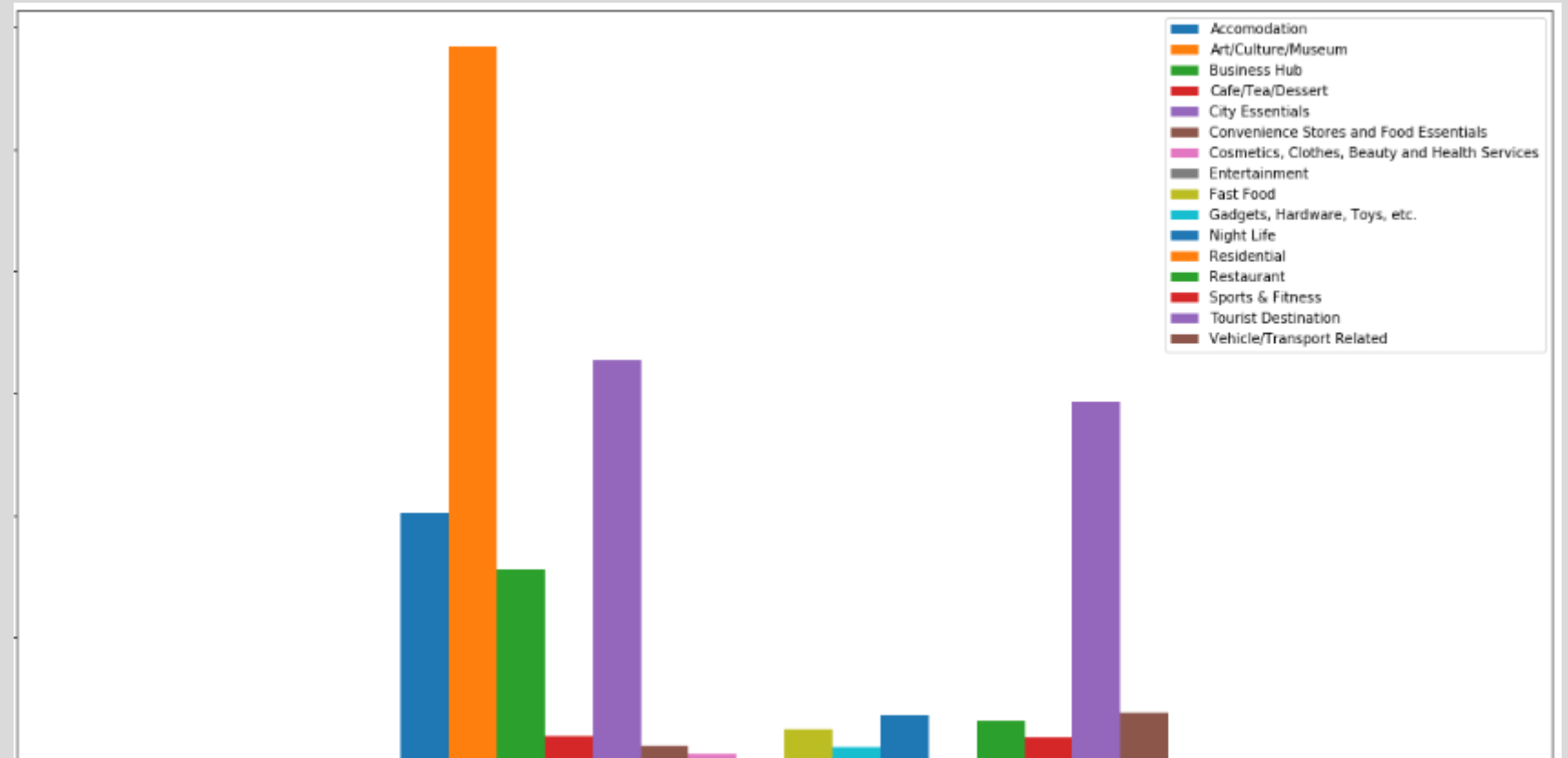


Most Optimal number  
of Cluster is 7

# Visualizing the Clusters

# Cluster 1: Arts, Culture and Science District

Notice the high value for Arts, Culture and Museum

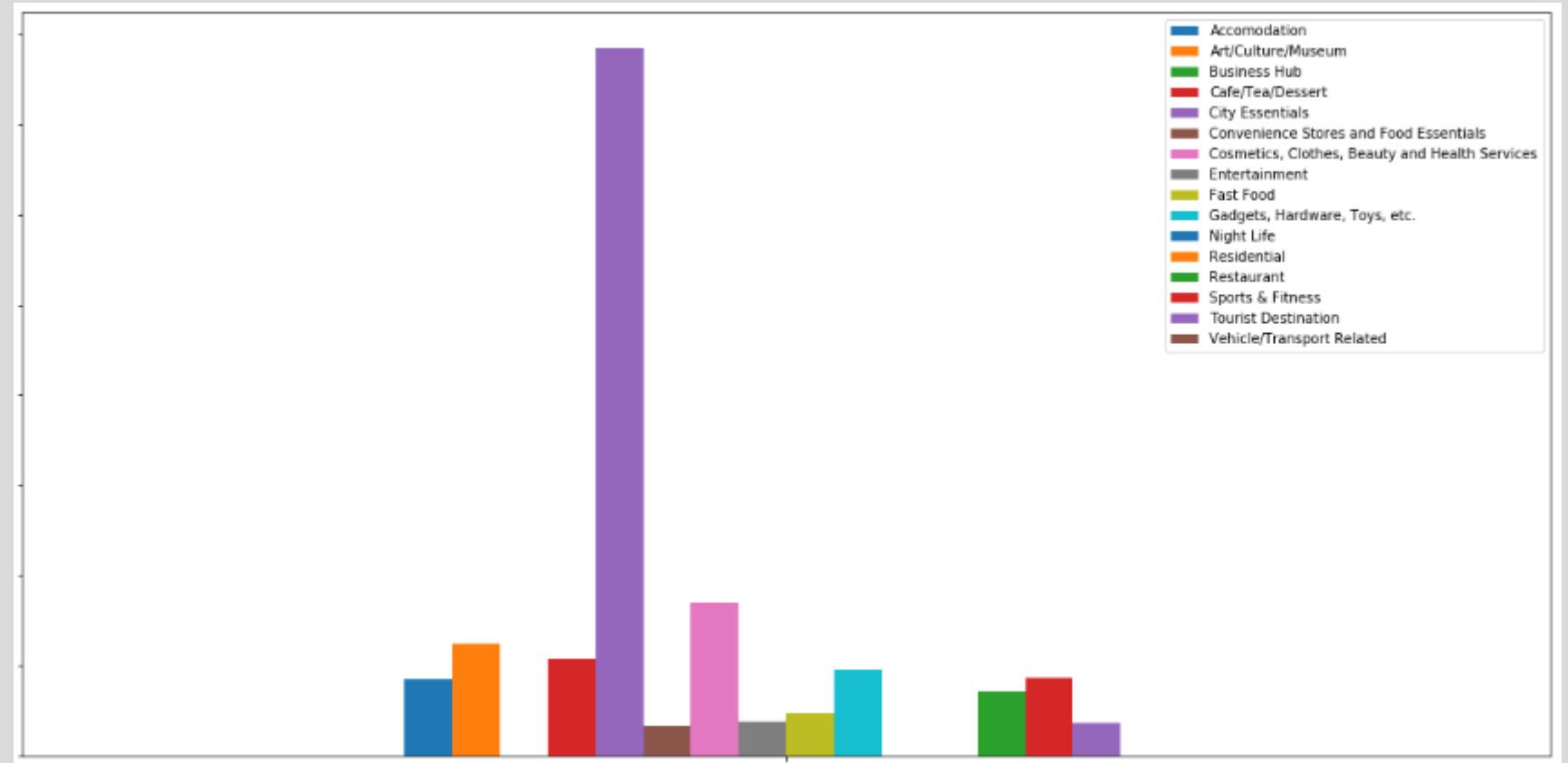


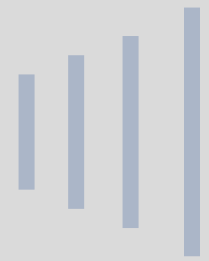


# Cluster 2: Downtown Church, Hospital or School Zone

Notice the high value for the city essentials bar

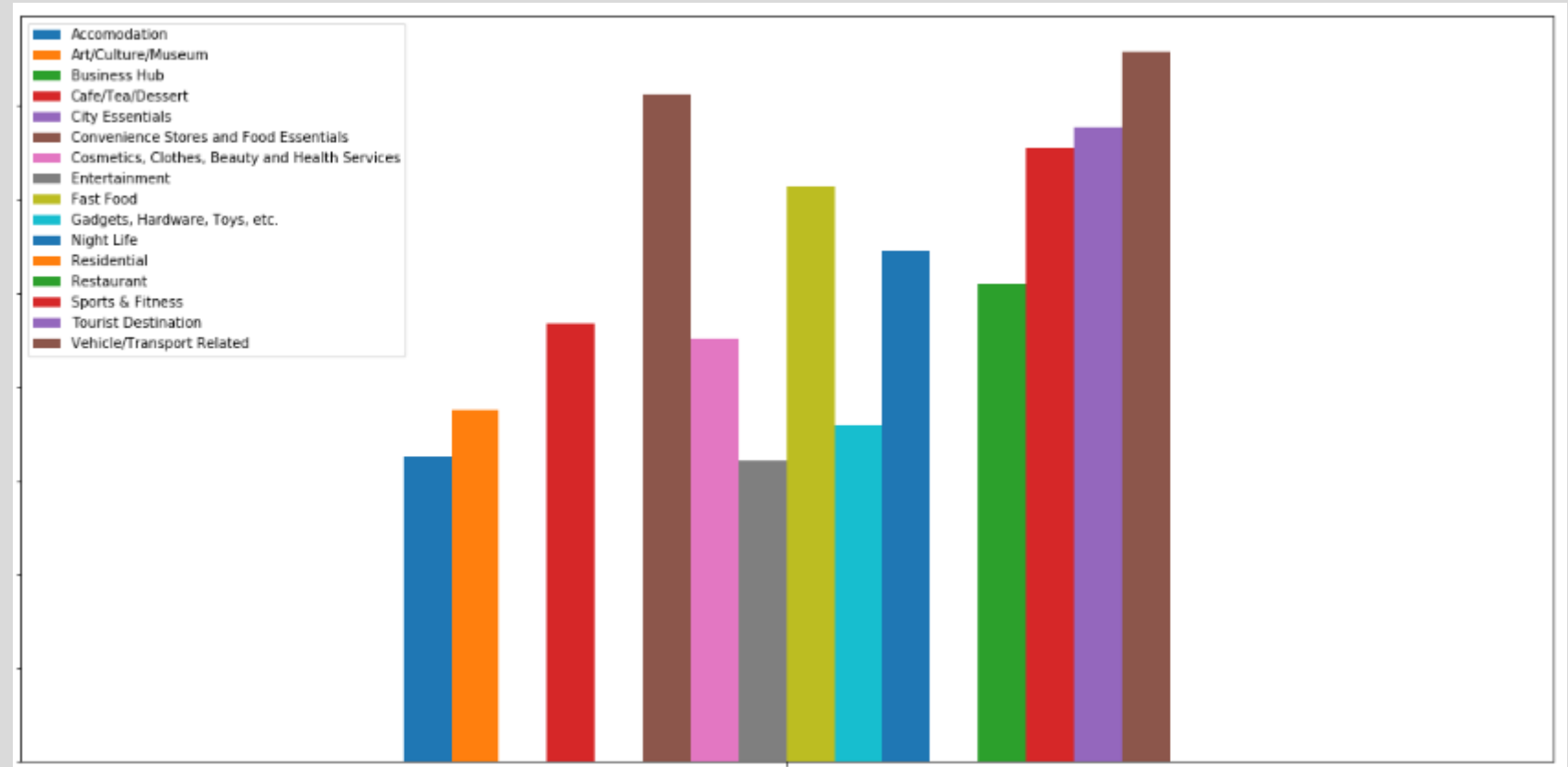
Compare this graph to Cluster 7 which will be discussed later





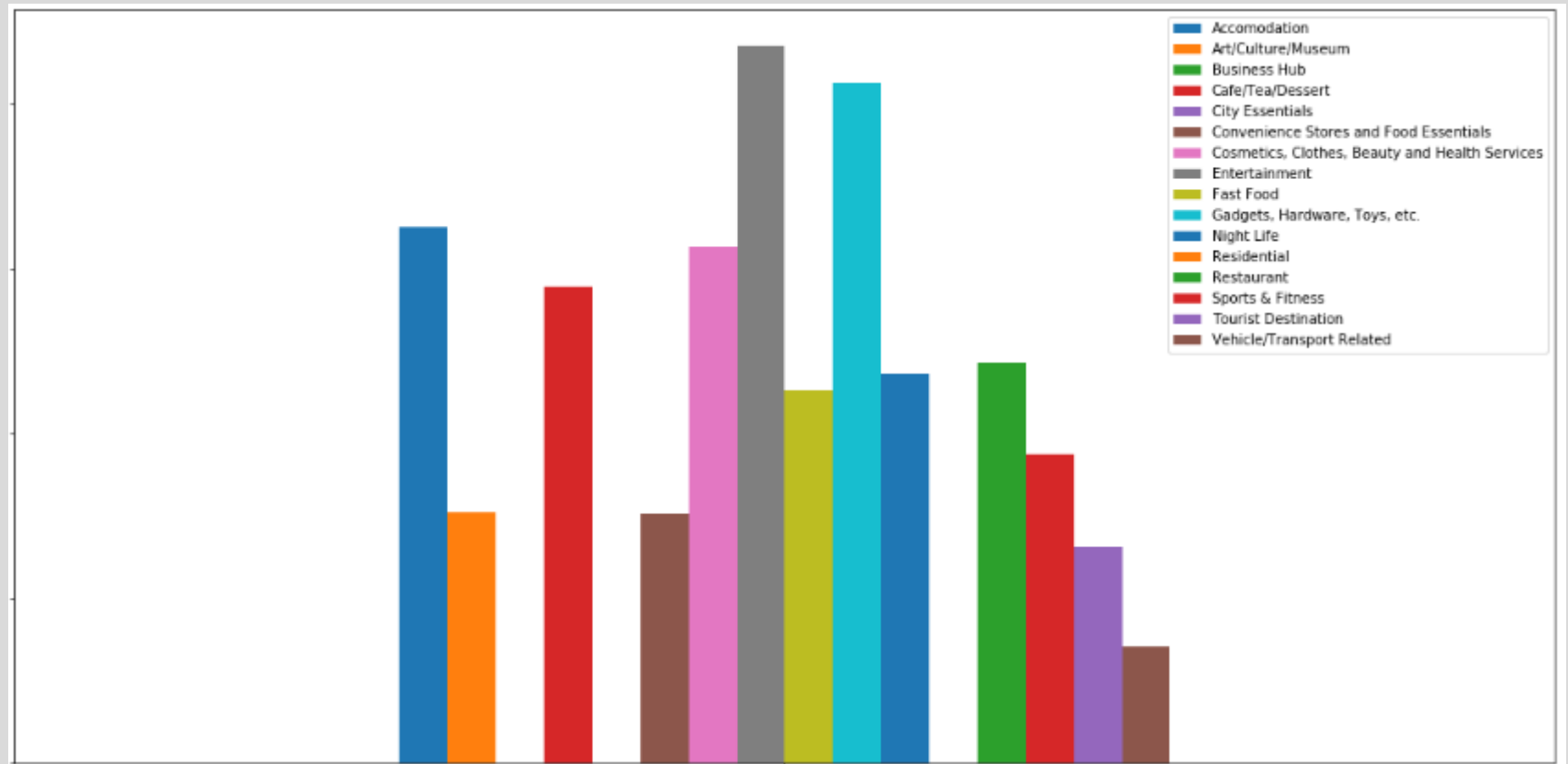
# Cluster 3: City Center

Notice the high value for most of the bars especially the Vehicle/Transport related and Convenience Stores and Food essential venue types



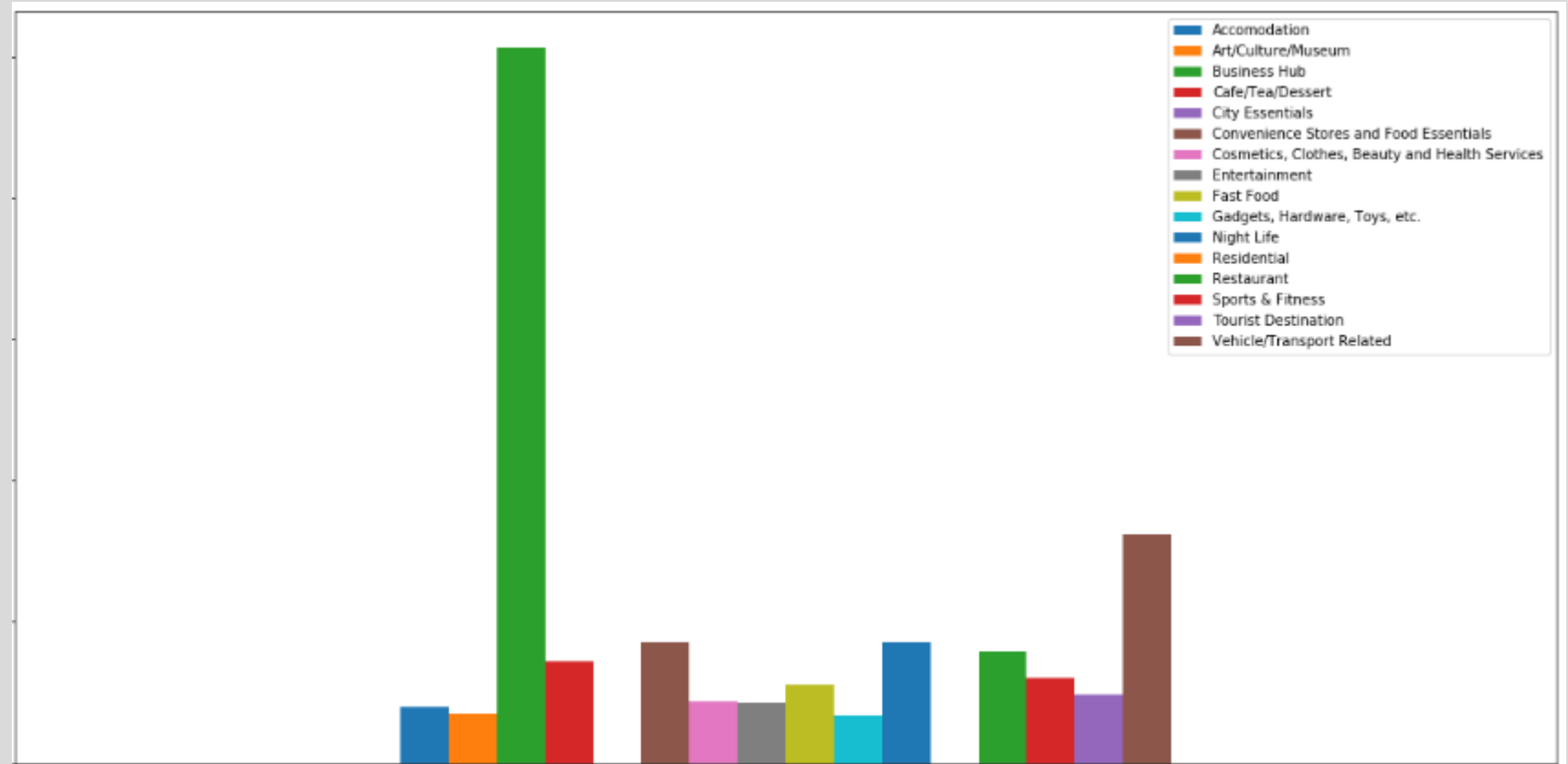
# Cluster 4: Entertainment and Shopping District

Notice the high value for Entertainment and Cosmetics, Clothes, Beauty and Health services venue types



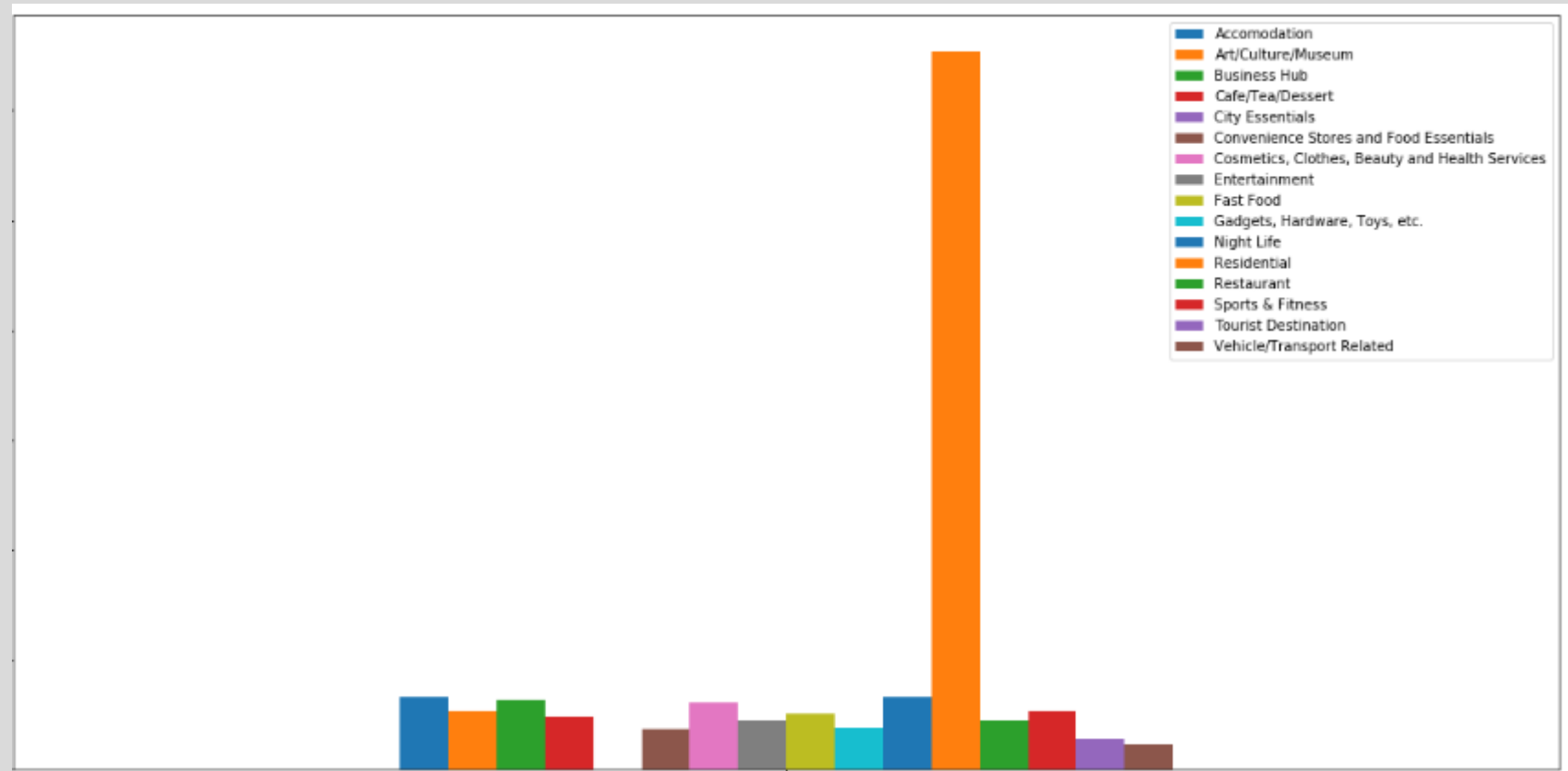
# Cluster 5: Business District

Notice the high value for Business Hub Venue types



# Cluster 6: Residential District

Notice the high value for residential venue types

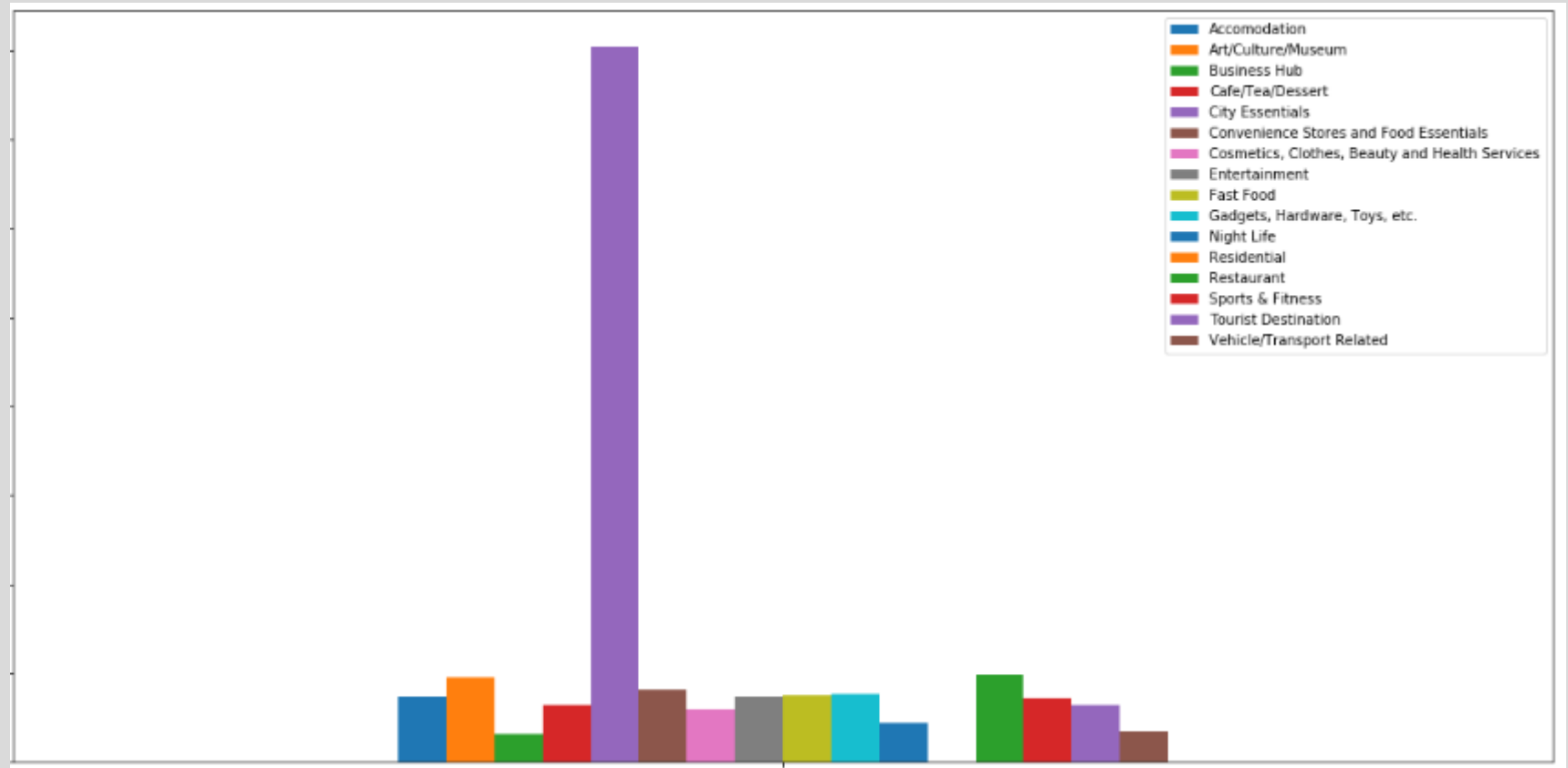




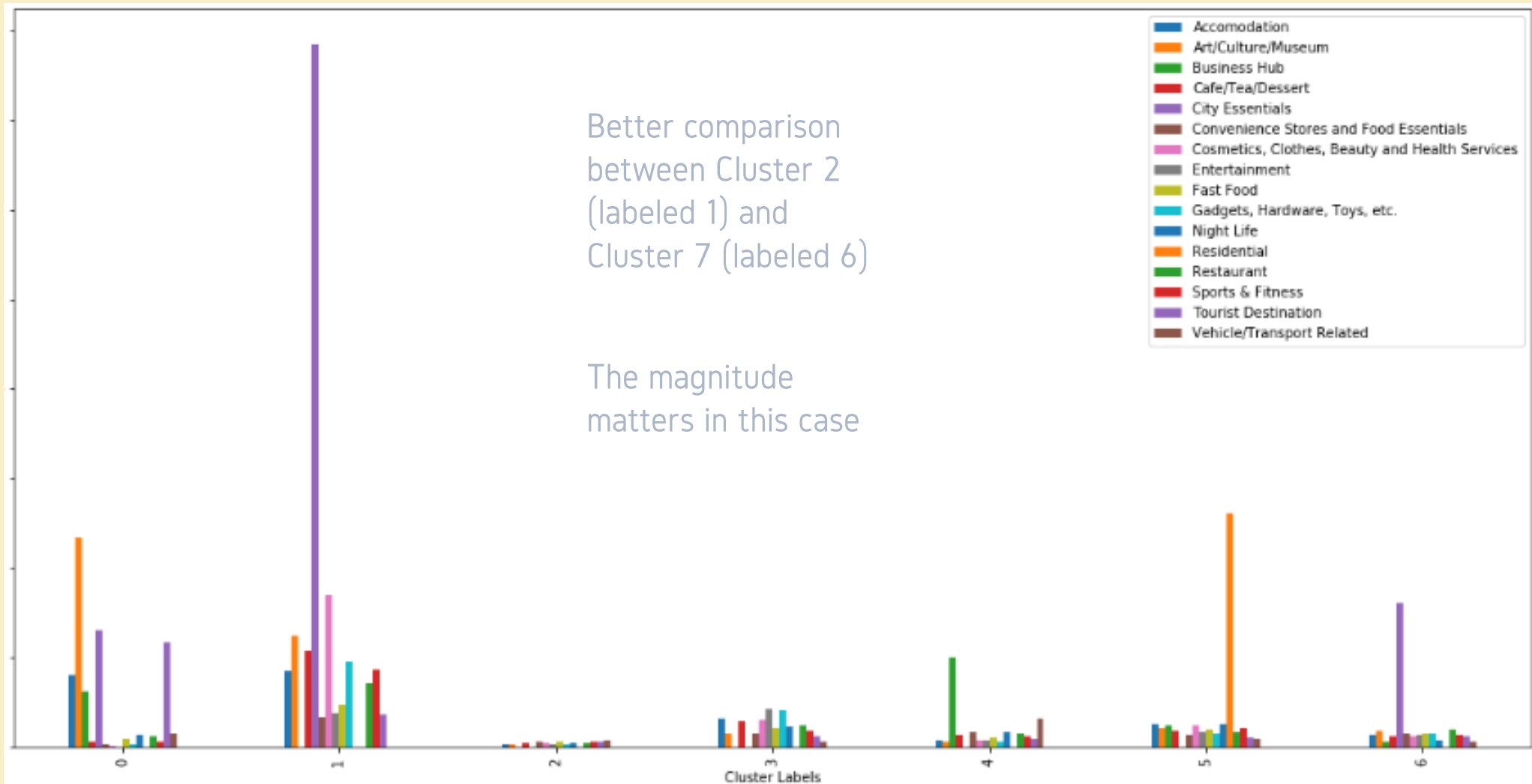
# Cluster 7: Low Profile Church, Hospital or School Zone

Notice the high value for city essential venue types

Smaller value of the other venue types especially compared to Cluster 2



# Bar graph of all the Clusters



# Conclusion

Neighborhoods in Metro Manila is divided into 7 clusters

The information can be used by business owners and people who plans to migrate to the different neighborhoods

| Cluster | Description                                 |
|---------|---|
| 1       | Arts, Culture and Science District          |
| 2       | Downtown Church, Hospital or School Zone    |
| 3       | City Center                                 |
| 4       | Entertainment and Shopping District         |
| 5       | Business District                           |
| 6       | Residential District                        |
| 7       | Low Profile Church, Hospital or School Zone |

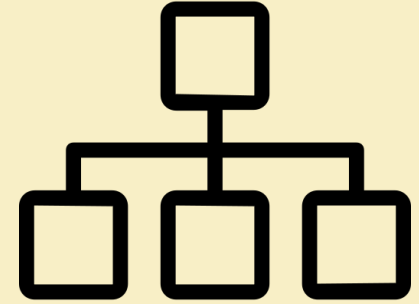
# Recommendations



- Add more datasets that would visualize the neighborhoods in Metro Manila better.
  - E.g. Real Estate Prices
  - Population Density and Demographics



- Use more official dataset other than scrapping it from websites



- Use other Clustering Models such as Hierarchical Clustering or DBSCAN Models