IBM DATA SCIENCE – CAPSTONE Battle Of The Neighborhoods

Choosing The Best Vacation Destination Between Dubai, Paris, New York and Kuala Lumpur

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Introduction and Problem Description

In this Capstone Project, I will use my learnings and understandings from IBM's Professional Data Science Certificate to solve this never-ending problem of which place is it better to go for a vacation between these four famous cities. Many foreign tourists are trying to find a vacation spot which gives the whole package and is worth them taking their time out of their lives to come and enjoy. Tourists want to choose a place that has a variety of activities to do and a variety of places to go. In this project I will try to explore the two cities to see which city offers the best vacation destination in terms of hotels, restaurants, malls, museums, sight-seeing venues and fun outdoor activities.

This data will be able to solve this problem as we can see what the reviews, ratings and tips for places in the city are. How many 5 – Star hotels does the city have. How many different categories of places are available and how far are their distances from the places where the tourists plan to live. Furthermore, we can see what Tourists look for in ideal vacation venues and cross check them with cities to see whether those cities provide those places or not.

Data Section

1. Requirements

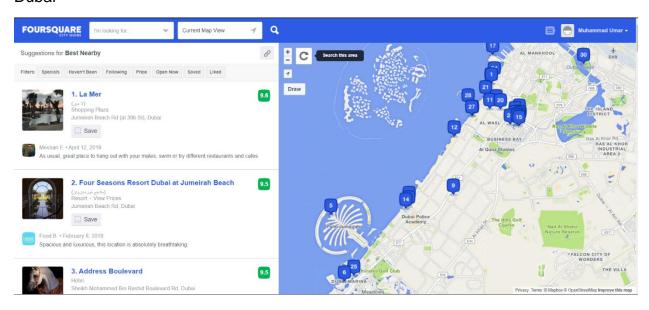
The data required in this is all types of categories for vocational purposes such as Hotels, Restaurants, Shopping Malls, Museums, Sight-Seeing Venues and Fun-Outdoor Activities. For that purpose, I will use the Foursquare API to 'Search' and 'Explore' all these Vacation venues to observe and do exploratory data analysis on the following cities.

2. Source

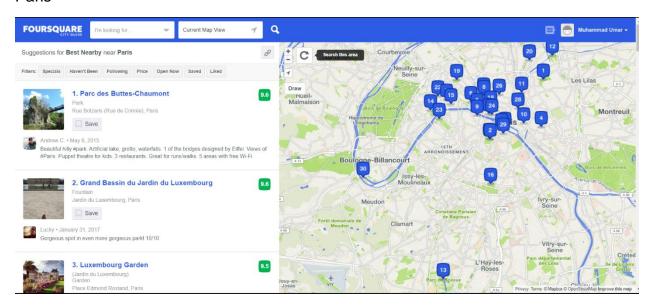
The data for this project will be collected using Foursquare API using the following link: https://foursquare.com/developers. In order to extract our data, I will use my Client ID and Client Secret to create a query which will fetch the data I need from the website.

```
1 CLIENT_ID = '******' # your Foursquare ID
2 CLIENT_SECRET = '******' # your Foursquare Secret
3 VERSION = '*****' # Foursquare API version
```

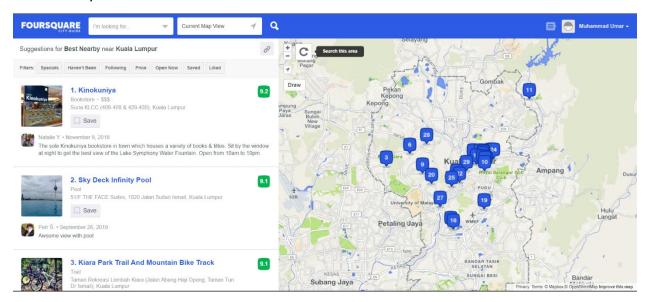
Dubai



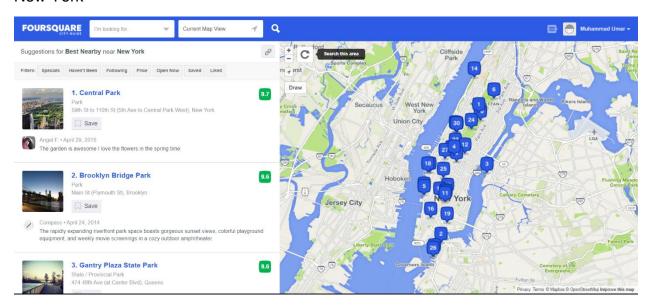
Paris



Kuala Lumpur



New York



3. Audience

The audience for this report will be the foreign tourists and travelers looking for their perfect vacation destination. It will not only benefit these travelers but will also give insight to the tourist industry as to what things make a place a better spot and more likely place for people to go for their holidays. **Tourism** industry is important for the benefits it brings and due to its role as a commercial activity that creates demand and growth for many more industries.

Methodology Section

1. Feature Engineering

First of all, we requested venues from the Foursquare API as shown above. For each of the 4 cities we looked at all the with a RADIUS of 10000. From the Geolocator we found out the location of all the cities and mapped them out as shown below.

Paris



Dubai



New York



Kuala Lumpur



We then converted the raw JSON to a DataFrame.

| | id | name | categories | referralld | hasPerk | location.address | location.lat | location.Ing | location.labeledLatLng |
|---|--------------------------|---|--|------------------|---------|---|--------------|--------------|--|
| 0 | 4c73295213228cfa74312d65 | Dataran DBKL | [{'id': '4bf58dd8d48988d163941735', 'name': 'P | V- 1585689957 | False | Menara DBKL | 3.151427 | 101.694462 | [{'label': 'display', 'la 3.151427172961877 |
| 1 | 4c75ac31ff1fb60ca9faf6a7 | Menara DBKL 1 | [{'id': '4bf58dd8d48988d129941735', 'name': 'C | V- 1585689957 | False | Jalan Raja Laut | 3.152314 | 101.694662 | [{'label': 'display', 'la 3.152314298073572 |
| 2 | 4ec4f9b9e300e6894f206ae2 | Dewan Bandaraya Kuala Lumpur Menara 2 | [{'id': '4bf58dd8d48988d129941735', 'name': 'C | V- 1585689957 | False | Bangunan DBKL Menara 2, Tingkat 6 | 3.151609 | 101.694551 | [{'label': 'display', 'la 3.15160866153506, |
| 3 | 4d65cd0a56746dcbae433fff | Cafeteria DBKL | [{'id': '4bf58dd8d48988d142941735', 'name': 'A | V- 1585689957 | False | Tkt 2, Menara DBKL 1 | 3.152154 | 101.694922 | [{'label': 'display', 'la 3.152153558414952 |

However, a lot of feature engineering had to be done. So firstly, we dropped all the columns which we were not going to need, and which

didn't give us any meaning. Then we looked into categories and extracted out the venue. We divided all the venues for all the city Data Frames into 6 very generic categories all which would be essential in a vacation trip. The following are:

| | Category | Rank |
|---|--------------|------|
| 0 | Hotels | 1 |
| 1 | Activities | 2 |
| 2 | Food | 3 |
| 3 | Sight Seeing | 4 |
| 4 | Travel | 5 |
| 5 | Shopping | 6 |

We assigned a rank to each of the categories based on the importance it had for tourists which on popular opinion was in this order. After all the Feature Engineering we managed to shape our Data Frame to be exactly as we had hoped.

| | id | name | categories | latitude | longitude | distance |
|----|--------------------------|--------------------------------|--------------------|-----------|------------|----------|
| 0 | 58f0cd8db1ec13241266420a | New York | Pizza Place | 40.712784 | -74.005940 | 8 |
| 1 | 4a676321f964a52051c91fe3 | New York City Hall | City Hall | 40.712659 | -74.005880 | 13 |
| 2 | 4b475390f964a520f12e26e3 | Mary's Coffee Shop | Coffee Shop | 40.712786 | -74.005944 | 8 |
| 3 | 4b57b0dff964a520293c28e3 | MTA Subway - City Hall (R/W) | Metro Station | 40.713394 | -74.006934 | 107 |
| 4 | 4b79a5e8f964a52037082fe3 | NY Gift Shop | Gift Shop | 40.712733 | -74.005978 | 3 |
| 5 | 3fd66200f964a520d8f11ee3 | City Hall Park | Park | 40.712415 | -74.006724 | 69 |
| 6 | 51a4bc7c498e469047be66d6 | City Hall Council Chambers | City Hall | 40.712736 | -74.005472 | 45 |
| 7 | 5ae3d3db3ba767002cf97a85 | El Mexicano Restaurante & Cafe | Mexican Restaurant | 40.675417 | -74.000486 | 4179 |
| 8 | 4dd77d22b3adc64ae076925d | St comunity beach | Beach | 40.712781 | -74.005944 | 8 |
| 9 | 51b3b7da498efda7cefb2660 | Dahlia's Mexican Restaurant | Mexican Restaurant | 40.718754 | -74.010337 | 763 |
| 10 | 4e31861d7d8b9b256bcc43a8 | June's Fruit Cart | Food Truck | 40.712786 | -74.005944 | 8 |

2. Data Analysis

We then created for all cities a single Data Frame which we were going to use as a metric to discover which city gave the best results. From all the categories returned for each city we first took a count of the categories which actually fit for our problem. Then we took counts for all the individual broad categories for every city. Since we ranked the categories in terms of importance, we then calculated the weighted count for each individual category. The constructed data frame is shown below:

| | City | Category Count | Hotels | Activities | Food | Sight Seeing | Travel | Shopping | Hotels Weighted | Activities Weighted | Food Weighted | Sight Seeing Weighted | Travel Weighted | Shopping Weighted | Weighted Category Count |
|---|-----------------|-------------------|--------|------------|------|-----------------|--------|----------|--------------------|------------------------|------------------|-----------------------------|--------------------|----------------------|-------------------------------|
| 0 | Paris | 26 | 1 | 2 | 11 | 2 | 4 | 6 | 6 | 10 | 44 | 6 | 8 | 6 | 80 |
| 1 | Kuala Lumpur | 38 | 2 | 7 | 18 | 6 | 0 | 5 | 12 | 35 | 72 | 18 | 0 | 5 | 142 |
| 2 | New York | 44 | 4 | 5 | 21 | 4 | 6 | 4 | 24 | 25 | 84 | 12 | 12 | 4 | 161 |
| 3 | Dubai | 33 | 1 | 18 | 10 | 2 | 0 | 2 | 6 | 90 | 40 | 6 | 0 | 2 | 144 |

We then marked the clusters for each category on the map. The diagrams are shown in the result section. We added up all the values for each city and got a total score. To normalize the score based on the number of categories that were actually present in that vicinity of the city we divided the Total by the Category count to get the normalized scores for each city. We sorted them by these values which led us to discover which city was the best in terms of the facilities they provide for tourists on vacations.

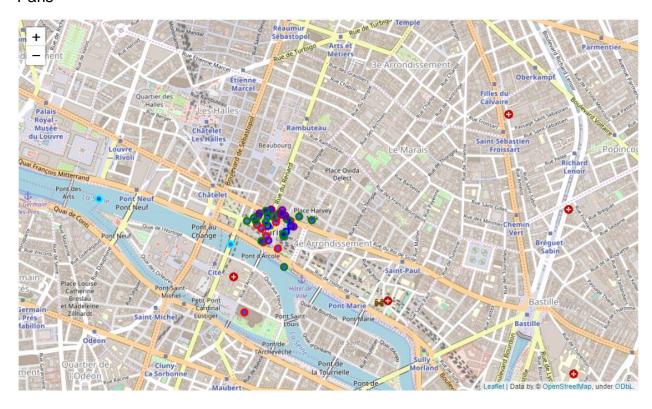
Results

First, we will see the clusters formed from the mapping of the categories in each city as show below:

Color of the clusters

- Hotels Yellow
- Activities Blue
- Food Green
- Sight Seeing Cyan
- Travel Red
- Shopping Purple

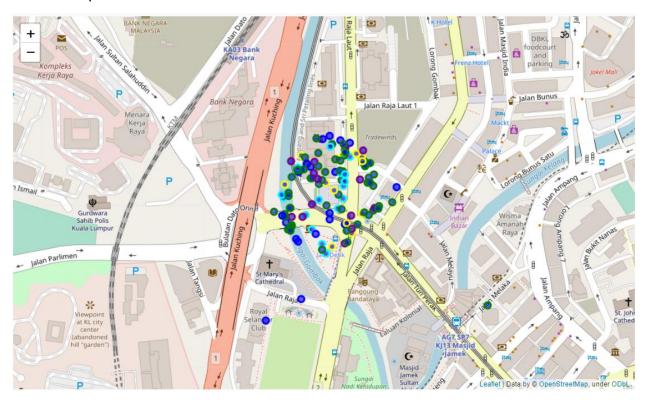
Paris



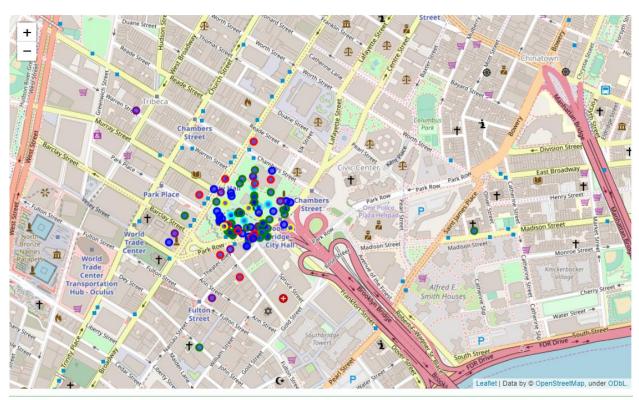
Dubai



Kuala Lumpur



New York



Now looking the total scores of each of the cities

| City | Category Count | Hotels | Activities | Food | Sight Seeing | Travel | Shopping | Hotels Weighted | Activities Weighted | Food Weighted | Sight Seeing Weighted | Travel Weighted | Shopping Weighted | Weighted Category Count | Total |
|-------------------|-------------------|--------|------------|------|-----------------|--------|----------|--------------------|------------------------|------------------|-----------------------------|--------------------|----------------------|-------------------------------|-------|
| 0 Paris | 26 | 1 | 2 | 11 | 2 | 4 | 6 | 6 | 10 | 44 | 6 | 8 | 6 | 80 | 212 |
| 1 Kuala Lumpur | 38 | 2 | 7 | 18 | 6 | 0 | 5 | 12 | 35 | 72 | 18 | 0 | 5 | 142 | 360 |
| 2 New York | 44 | 4 | 5 | 21 | 4 | 6 | 4 | 24 | 25 | 84 | 12 | 12 | 4 | 161 | 410 |
| 3 Dubai | 33 | 1 | 18 | 10 | 2 | 0 | 2 | 6 | 90 | 40 | 6 | 0 | 2 | 144 | 354 |

We then normalized these results and got the following

| City | Score |
|--------------|-----------|
| Dubai | 10.727273 |
| Kuala Lumpur | 9.473684 |
| New York | 9.318182 |
| Paris | 8.153846 |

From the following results we will now draw our conclusion discuss further observation that could be done.

Discussion

From the results above and the exploratory data analysis we can see that further analysis can be done by considering tips and ratings of all the venues. We could also look at successful countries in vacations and do a cluster analysis and see as to which place resembles the target place which is known to be a good vacation spot. We could also read reviews and use Natural language processing to see whether the reviews are positive or negative sentiment

Conclusion

From the following report and analysis that we have conducted it is the results show that Dubai among Paris, New York and Kuala Lumpur is the vacation destination as it scored most in all relevant categories and had the highest score. Due to Dubai's many outdoor activities and 5 star hotels and amazing Metro travel station it is clear that it deserves first place.