

# Analyzing City Specific Data to Predict Hotel Density around Airports

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# Problem Statement

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A travel agency which mainly deals with hotel bookings has multiple locations across the United States is expanding. The management team is looking into the options and trying to decide which city would be best suitable to target for business opportunities.

Good starting point is to explore the cities which have following characteristics:

- ☐ City is a commercial hub
- ☐ City has tourist attractions
- ☐ City has airports
- ☐ City has big volume of business and leisure travelers

# Data Used

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- ❑ Annual enplanement information of Top 30 US airports
  - ❑ Scraped from Wikipedia
- ❑ GDP information corresponding to cities with high annual enplanements
  - ❑ Scraped from Wikipedia
- ❑ Number of Hotels/Motels/Resorts around the airport.
  - ❑ Fetched using Foursquare API

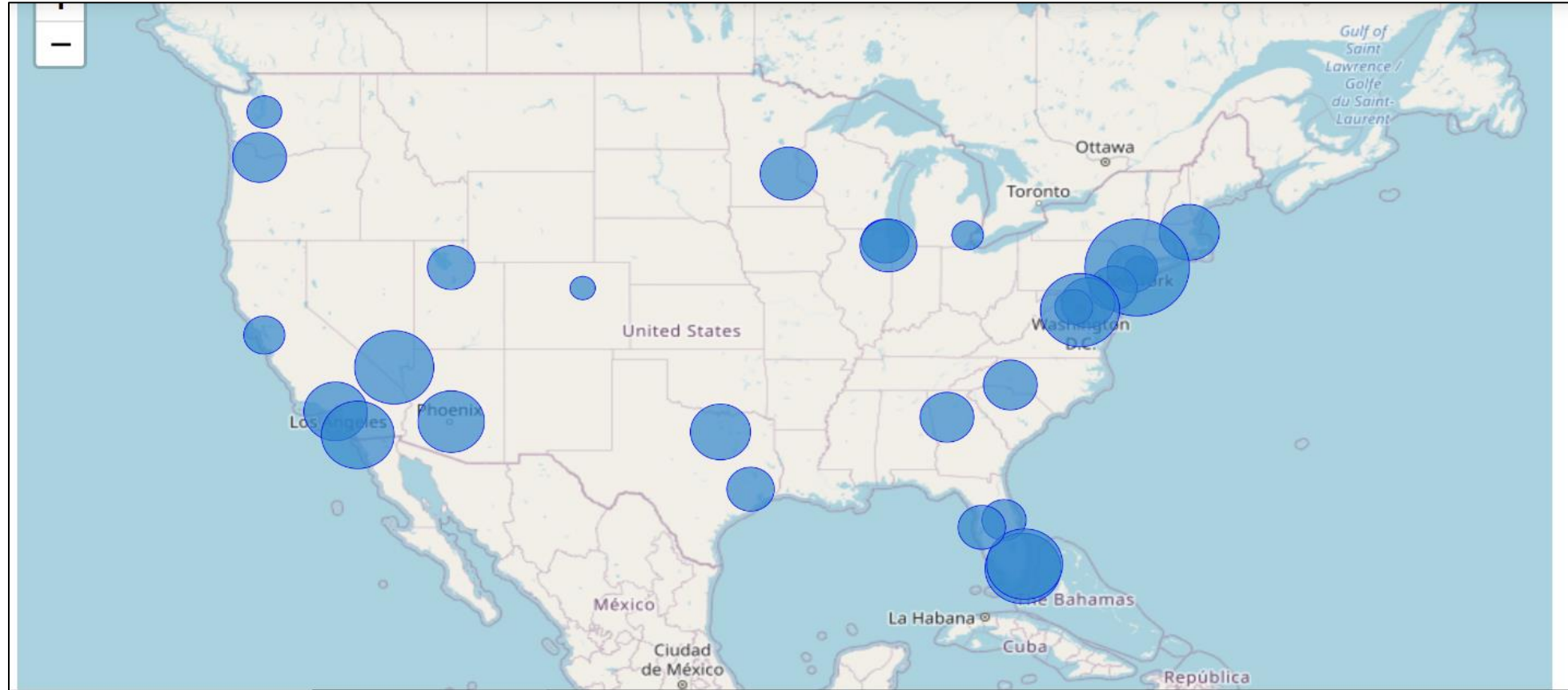
# Top Rows of Data Frame Used in Analysis

```
gdp_airport = gdp_airport.join(top_airports['IATACode'])
gdp_airport
```

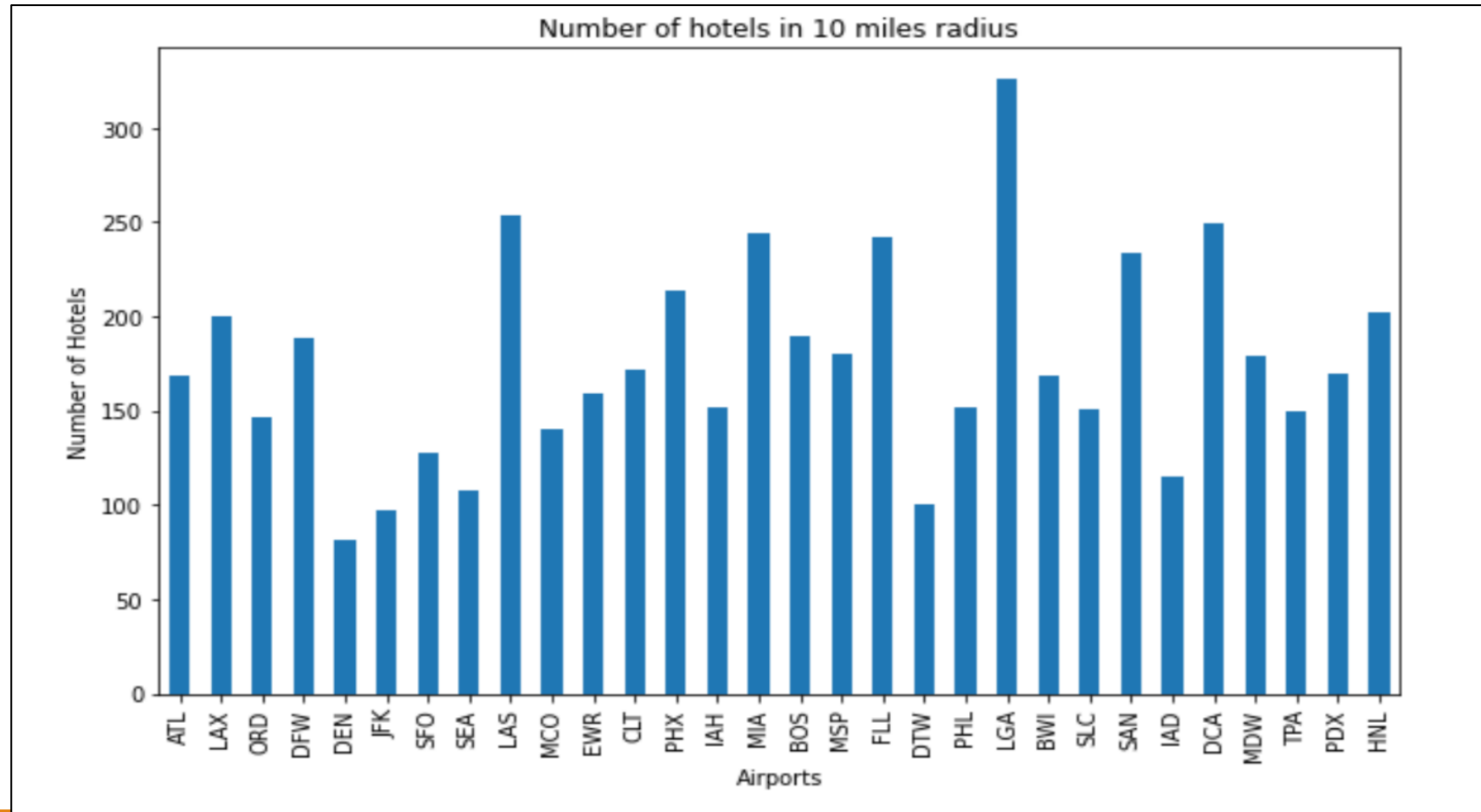
Out[24]:

|   | Airport  | City          | Area  | AirTraffic | GDP     | Latitude  | Longitude   | NumHotels | IATACode |
|---|--|---------------|---|------------|---------|-----------|-------------|-----------|----------|
| 0 | Hartsfield–Jackson Atlanta International Airport | Atlanta       | Atlanta-Sandy Springs-Roswell, GA (Metropolita... | 51866464   | 397261  | 33.637799 | -84.429271  | 169       | ATL      |
| 1 | Los Angeles International Airport                | Los Angeles   | Los Angeles-Long Beach-Anaheim, CA (Metropolit... | 42626783   | 1047661 | 33.942168 | -118.421376 | 200       | LAX      |
| 2 | O'Hare International Airport                     | Chicago       | Chicago-Naperville-Elgin, IL-IN-WI (Metropolit... | 39874879   | 689464  | 41.977985 | -87.909321  | 146       | ORD      |
| 3 | Dallas/Fort Worth International Airport          | Dallas        | Dallas-Fort Worth-Arlington, TX (Metropolitan ... | 32800721   | 512509  | 32.896519 | -97.046522  | 188       | DFW      |
| 4 | Denver International Airport                     | Denver        | Denver-Aurora-Lakewood, CO (Metropolitan Stati... | 31363573   | 214157  | 39.850188 | -104.675328 | 81        | DEN      |
| 5 | John F. Kennedy International Airport            | New York      | New York-Newark-Jersey City, NY-NJ-PA (Metropo... | 30541459   | 1772319 | 40.642948 | -73.779373  | 97        | JFK      |
| 6 | San Francisco International Airport              | San Francisco | San Francisco-Oakland-Berkeley, CA (Metropolit... | 27794154   | 548613  | 37.622452 | -122.384072 | 128       | SFO      |
| 7 | Seattle–Tacoma International Airport             | Seattle       | Seattle-Tacoma-Bellevue, WA (Metropolitan Stat... | 24894338   | 392036  | 47.447567 | -122.308016 | 108       | SEA      |
| 8 | McCarran International Airport                   | Las Vegas     | Las Vegas-Henderson-Paradise, NV                  | 22955085   | 100400  | 36.084188 | -115.138188 | 854       | LAS      |

# Airport Vs Number of Hotels in 10 Miles Radius on a Map

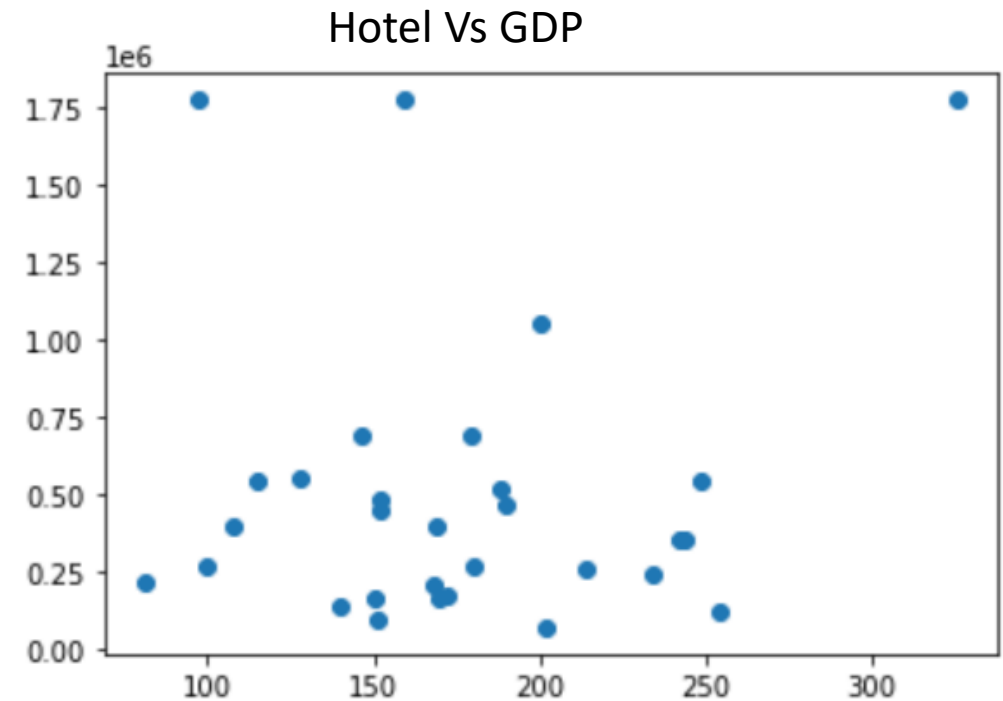
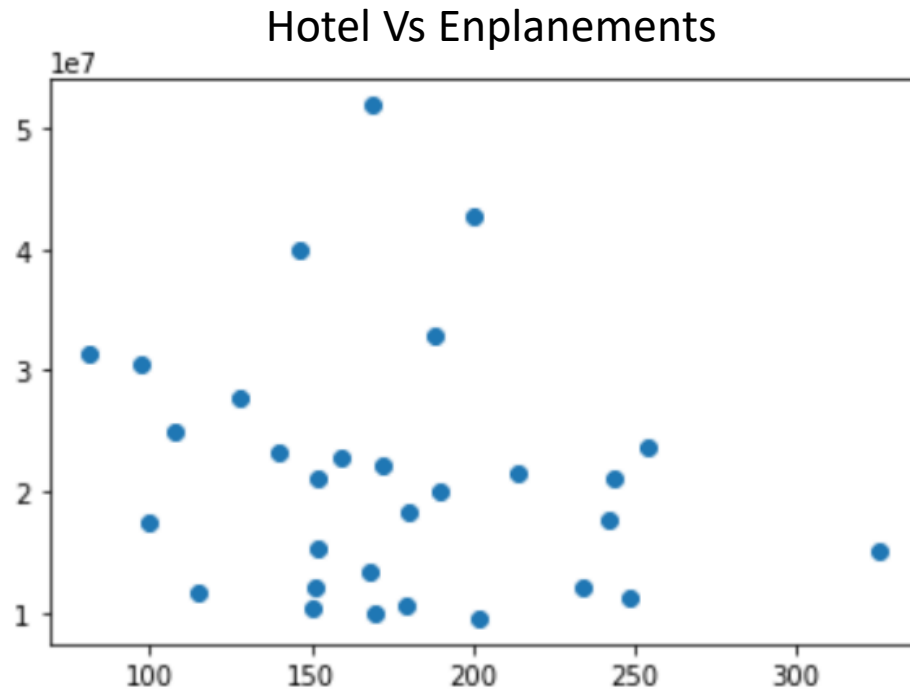


# Airport Vs Number of Hotels in 10 Miles Radius – Bar Plot



# Building Multi Linear Regression Model – Scatter Plots of Independent Vs. Dependent Variables

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# Building Multi Linear Regression Model

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```
▶ X = pd.DataFrame([gdp_airport['GDP'], gdp_airport['AirTraffic']])  
X = X.T  
y = pd.DataFrame([gdp_airport['NumHotels']])  
y = y.T  
  
lm = LinearRegression()  
lm.fit(X,y)  
lm.score(X,y)
```

```
2]: 0.0747813627408721
```



# Results

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It can be seen that  $R^2$  value is pretty low in this case which confirms that the relation between dependent and independent variables is not linear at all.

The analysis of Top-30 US airports shows that GDP & Annual Enplanements is not a good measure for desired prediction. This could be explained by following reasons:

- ❑ In some cities the nearest major airports are within the city limits and in others they are in outskirts. If airport is in outskirts then hotels are probably going to be far from the airport even if the airport is busy and city has strong GDP.
- ❑ The high enplanement rates could also mean that the airport is only serving as a hub, and a lot of passenger traffic is only connecting passengers.
- ❑ Sometimes a single city can have multiple airports. In our data set New York has two major airports; John F. Kennedy (JFK) and La Guardia (LGA). JFK has high annual enplanements as compared to LGA but Foursquare data shows that LGA has 326 hotels around it while JFK only has 97.

# Conclusion

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The travel agency should explore other criteria like number of big companies in the area, employment rates of the city, GDP per capita, major tourist attractions, number of universities, conference centers etc. to be used as good predictors to estimate hotel density around any airport.