# Analysis of tourist arrivals in cities around the world



- City tourist management agencies can be very interested to have a tool that they can use to increase the number of annual tourists and improve the economy of their city.
- The tool can be used to understand which services need to be improved to attract more tourists

## 2. Data acquisition and cleaning

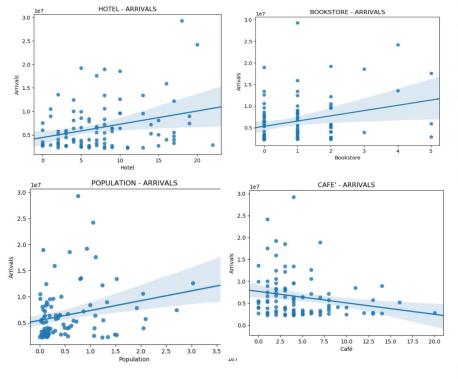
### DATA ACQUISITION:

- LIST OF CITIES 1: I use the "List of cities by international visitors" (https://en.wikipedia.org/wiki/List\_of\_cities\_by\_international\_visitors) classified by the Euromonitor Rank. I'm intersted in the "City" and "2018 Arrivals" columns. For evey city i will find Latitude and Longitude with geolocator and add this two columns to the DataFrame.
- LIST OF CITIES 2: I use the list of city by population (csv file: https://worldpopulationreview.com/world-cities)
- LIST OF CITIES 3: I use the list of city by area (http://www.citymayors.com/statistics/largest-cities-area-125.html).
- I merge the three data-frame. The Data frame will have the following columns: City, Population, Area, Arrivals, Latitude, Longitude, Population, Land Area, Arrivals.
- FINAL DATASET: I will add to this dataset the number of venues (restaurant, hotel, etc.) for each city, getting the data from Foursquare.

### DATA CLEANING AND FEATURES SELECTION:

Data downloaded or scraped from multiple sources were combined into one table. There were a lot of missing values, mainly geographic and demographic data. I have to fill the missing values manually. After data cleaning and feature selection there were 100 samples and 28 features in the data.

### PEARSON CORRELATION AND P-VALUE



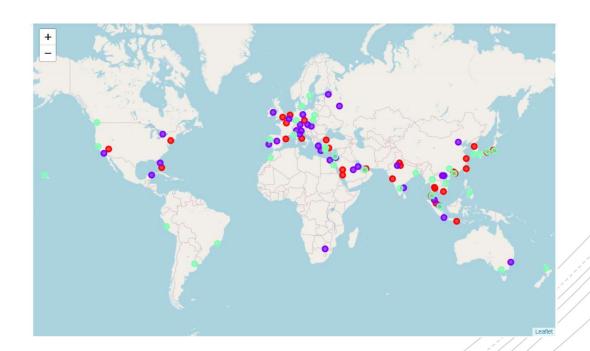
Pearson correlation and P-value

Venues	Corr arrivals	p_value arrivals
Café	-0.203843	0.0419
Burger Joint	-0.179115	0.0746
Pizza Place	-0.17905	0.0747
Restaurant	-0.137721	0.1718
Beach	-0.136569	0.1755
Resort	-0.130462	0.1958
Coffee Shop	-0.115496	0.2525
Supermarket	-0.07001	0.4888
Bakery	-0.0614444	0.5437
Bar	-0.0459773	0.6497
Plaza	-0.0231188	0.8194
Land Area	-0.0216751	0.8305
Scenic Lookout	-0.0166459	0.8694
Theater	-0.0121452	0.9045
Ice Cream Shop	-0.0117868	0.9073
Historic Site	-0.0102916	0.9191
Sandwich Place	-0.00861017	0.9322
Wine Bar	0.00172134	0.9864
Dessert Shop	0.00972386	0.9235
Cocktail Bar	0.0331894	0.7431
Park	0.041282	0.6834
Art Museum	0.0525708	0.6034
Garden	0.17683	0.0784
Shopping Mall	0.188032	0.061
Population	0.261778	0.0085
Bookstore	0.28316	0.0043
Hotel	0.30597	0.002

3. Data Analysis

Cluster	Population	Land Area	Arrivals
High	8.546208e+06	1555.969697	1.196490e+07
Medium	4.610400e+06	1311.151515	4.971482e+06
Low	3 237467e+06	1850 882353	2 695406e+06

3. Cluster Creation



# 4. Predictive Model

### DECISION TREE:

Decision tree parameters:

```
DecisionTreeClassifier(class_weight=None, criterion='entropy', max_depth=6, max_features=None, max_leaf_nodes=20, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, presort=False, random_state=None, splitter='best')
```

- Model Evaluation: accuracy = 0,89
- Feature importance

Feauture	Ranking
Land Area	0.197321
Restaurant	0.175164
Coffee Shop	0.172012
Café	0.142952
Hotel	0.102380
Plaza	0.050713
Historic Site	0.049559
Theater	0.048597
Art Museum	0.034768
Garden	0.026535

5. Conclusions

Excluding the land area, this study shows that services (hotels, restaurants, etc.) are the structures that have the most influence on tourism numbers. The tourism management agencies they must therefore take into account that more than the attractions of the place (parks, historical sites, etc.) are not enough to increase the number of tourists, but they need to improve the services.