

IBM – Coursera
Data Science Specialization

Capstone project - Final report

**Analysing and comparing student friendly
neighbourhoods in USA**

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I. Introduction:

The students (all graduate levels) who decides to go to a university strive to know the various characteristics and benefits of the city/location. Both international and citizens search for common parameters in the city/state:

1. Housing
2. Transport facilities (public transport), magnitude of traffic incidents (crash, weather)
3. Corporate companies and tech startups (internship/full time)
4. Sports and recreation
5. Restaurants, malls, music concerts and game night pubs/bars
6. Libraries, medical centers and more

In this project, I try to solve this problem using foursquare location data and machine learning algorithms to compare and find the best places for students to find universities and settle in.

Stakeholders

This project will be useful to students, parents and graduate students (job seekers). This will be useful to other third parties who want to see the behaviour and psychology behind students' requirements that must be around the universities after learning what clusters usually those people fall into.

II. Data description:

We pull location/venue data that belong to above categories from foursquare and traffic incidents data from the selected state website. In this project, we consider only one state, New York city and its neighborhoods.

I obtain Incident/crash data from [New York Motor Vehicle Collisions – Crashes dataset](#). The dataset contains a vast number of features like latitude, longitude, number of persons injured/killed, number of pedestrians/cyclists injured, contributing factor vehicle 1 and more. But we just count the accidents per location or borough and append it to the dataset.

Regarding the venue data, it is same as what we saw in our previous exercise (Week 3). I will extract locations like universities, tech startups/corporate companies, restaurants/pubs, medical centers, events, etc. And, combine with the crash data.

For example, we extract venues using foursquare API: <https://api.foursquare.com/v2/venues/explore>. We get the data, normalize, find the frequency and append to the dataset.

Below is the example of the resultant data.

| | neighborhood | latitude | longitude | Medical_centers | Corporate_companies | Sports_venues | Restaurants | Libraries | Accidents |
|---|--------------|-----------|-----------|-----------------|---------------------|---------------|-------------|-----------|-----------|
| 0 | Marble Hill | 40.876551 | -73.91066 | 15 | 25 | 15 | 150 | 5 | 25 |
| 1 | Carnegie | 40.7845 | -73.9551 | 12 | 15 | 10 | 155 | 6 | 21 |

Also, this is a sample dataset for now. Dataset features can get modified as the project gets progressed.