### Introduction

New York City is a collection of many neighbourhoods scattered among the city's five boroughs—Manhattan, Brooklyn, the Bronx, Queens, and Staten Island—each exhibiting its own lifestyle. It is the most populous and the most international city in the country. Furthermore, New York is the most ethnically diverse, religiously varied, commercially driven, famously congested, and, in the eyes of many, the most attractive urban centre in the country.

Hence, opening any new business should be carefully studied and a strategic plan should be developed in order to guarantee the long term succes. Indeed, this requires a careful understanding of the diffrent neighbourhoods of the city, the ethnicity of the people and most important their habits and tastes.

# Business problem

We can say that healthcare is one of the most valuable industries in the World. This is due to the spread of the fast food culture which leads to many diseases and health problems. Hence, many people are trying to find methods in order to avoid falling into these problems. We can say that the preventive cure exists in two major categories: gym and healthy food.

In terms of healthy food, vegetrian restaurants come in the top of the list. Starting a vegterian restaurant can be a great business opportunity, but we need to distinguish ourselves from others to enjoy long-term success. We think that by exploring the neighbourhoods and their different venues and by emphesizing on the locations where gym centers are popular will lead us to the right place of our presumed vegetarian restaurants. This is because people who frequently visit gym are likely interested in eating healthy food.

We will content ourselves to provide this study for the city of New York and in particular to Manhattan neighbourhood since our client is interested in this area.

#### Data

We will rely on the publicly available data of New York city available online: <a href="https://geo.nyu.edu/catalog/nyu">https://geo.nyu.edu/catalog/nyu</a> 2451 34572

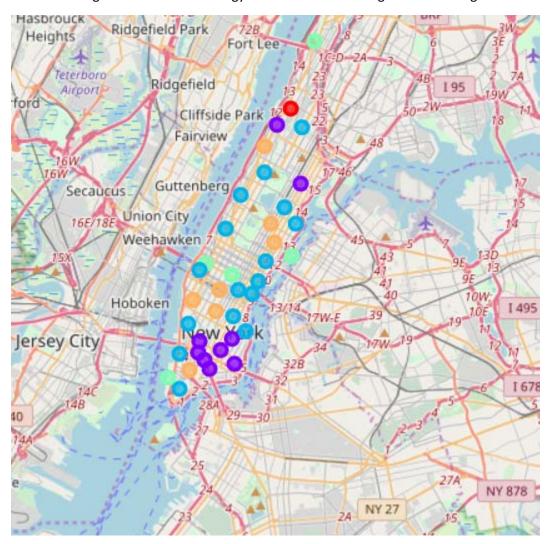
This dataset contains 5 boroughs and 306 neighborhoods. Now since our methodology is to search for gym/fitness center venues and try to plug our vegeterian restaurant nearby, we will use this website: <a href="https://developer.foursquare.com/docs/resources/categories">https://developer.foursquare.com/docs/resources/categories</a> in order to find gym/fitness center category id. We found that this id = 4bf58dd8d48988d175941735

# Methodology

Our methodology consists of loading and cleaning the dataset, then performing k-means statisical analysis in two stages. First, we consider venues with category = gym/fitness. Second, we do clustering regardless of the category type. Then we decide if a neighborhood needs a vegetarian restaurant, by inspecting if it appears in a cluster with high concentation in gym/fitness venues (i.e. by looking for clustering in Stage 1) and low concentration in restaurants, especially vegetarian restaurants (i.e. by looking for clustering in Stage 2).

## Results

After following the above methodology we obtained clustering in the first stage as shown below:



Now let's consider some of the resulting clusters. In particular let's take cluster 5:

#### Cluster 5 [34]: d.loc[manhattan\_merged['Cluster Labels'] == 4, manhattan\_merged.columns[[1] + list(range(5, manhattan\_merged t[34]: 1st Most 2nd Most 3rd Most 4th Most 5th Most 6th Most 7th Most 8th Most 9th Most Neighborhood Common Venue Common Venue Common Venue Common Venue Common Venue Common Common Common Venue Venue Venue Venue Gym / Doctor's Women's Martial Arts 8 Upper East Side Fitness Center Yoga Studio Pilates Studio Cycle Studio Medical Center Gym Office Store Dojo Gym / Doctor's 10 Lenox Hill Gym Pilates Studio Non-Profit Dance Studio Fitness Yoga Studio Cycle Studio Spa Office Gym / Martial Arts Recreation Gymnastics 17 Chelsea Fitness Yoga Studio Gym Cycle Studio Climbing Gym Pilates Studio Dojo Gym Center Center Gym / Morningside Basketball Cycle Studio 26 Gym Pool Boxing Gym Fitness Track Gym Yoga Studio Bridge Heights Court Center Gym / Martial Arts Athletics & Women's 32 Civic Center Fitness Pilates Studio Gym Pool Gym Yoga Studio Boxing Gym Dojo Sports Store Gym / Martial Arts Physical Fitness 33 Midtown South Gym Yoga Studio Pilates Studio Cycle Studio Building Boxing Gym Therapist

Clearly, this cluster contains plenty of gym/fitness centers.

Considering one of the neighborhoods in this cluster, for example 'Chelsea', and looking for it in the second stage of clustering, we got the following:

#### **Cluster 3**

| [5: | 1]: t   |                     | ged_all['C                              | luster Lab                              | els'] 2, m  | nanhattan_r                                       | merged_all  | .columns[[  | 1] + list                   | (range(5,                   | manhattan                   | _merged_a                   |
|-----|---------|---------------------|---|---|---|---|---|---|-----------------------------|-----------------------------|-----------------------------|-----------------------------|
| [5: | 1]:     | Neighborhood        |   | 1st Most<br>Common<br>Venue             | 2nd Most<br>Common<br>Venue                       | 3rd Most<br>Common<br>Venue                       | 4th Most<br>Common<br>Venue                       | 5th Most<br>Common<br>Venue                       | 6th Most<br>Common<br>Venue | 7th Most<br>Common<br>Venue | 8th Most<br>Common<br>Venue | 9th Most<br>Common<br>Venue |
|     | -       | 0                   | Marble Hill                             | Deli /<br>Bodega                        | Salon /<br>Barbershop                             | Residential<br>Building<br>(Apartment /<br>Condo) | High School                                       | Laundry<br>Service                                | Church                      | Furniture /<br>Home Store   | Nail Salon                  | Offic                       |
|     |         | 1                   | Chinatown                               | Chinese<br>Restaurant                   | Miscellaneous<br>Shop                             | Bus Station                                       | Bakery  | Bridge  | Noodle<br>House             | Park                        | Arts &<br>Crafts Store      | Bus Lin                     |
|     |         | 3                   | Inwood                                  | Laundry<br>Service                      | Deli / Bodega                                     | Salon /<br>Barbershop                             | Nail Salon  | American<br>Restaurant                            | Bank                        | Mexican<br>Restaurant       | Pharmacy                    | Bus Lin                     |
|     |         | 4                   | Hamilton<br>Heights                     | Residential<br>Building<br>(Apartment / | Salon /<br>Barbershop                             | Building  | Deli /<br>Bodega                                  | Non-Profit  | Laundry<br>Service          | Church                      | Bar                         | Bus Statio                  |
|     | Lincoli | n Square            | Opera<br>House                          |   | ool Theater                                       | Art Gallery                                       | Residential<br>Building<br>(Apartment /<br>Condo) | Building  | Performing<br>Arts Venue    | Library                     | Event<br>Space              | 0                           |
|     |         | Clinton             | Building                                | Bus Statio                              | on Theater  | Bus Line  | Lounge  | Residential<br>Building<br>(Apartment /<br>Condo) | Restaurant                  | Gym /<br>Fitness<br>Center  | Deli /<br>Bodega            | Cocktail                    |
|     |         | Chelsea             | High School                             | Buildir                                 | Residential<br>Building<br>(Apartment /<br>Condo) | Deli /<br>Bodega                                  | Office  | Art Gallery                                       | Pet Store                   | Pharmacy                    | Music<br>Venue              | Laui<br>Ser                 |
|     | G       | reenwich<br>Village | Residential<br>Building<br>(Apartment / |   |   | Furniture /<br>Home Store                         | Cocktail Bar                                      | Jewelry<br>Store                                  | Kids Store                  | Tech<br>Startup             | Ice Cream<br>Shop           | Clothing S                  |

We observe that Chelsea falls into cluster 3 which lacks of restaurants offering healthy food in general.

Hence we can consider that Chelsea is a good candidate for opening a new Vegetarian restaurant.

### Discussion

Clearly, the methodology that we followed in this project can be beneficial for stakeholders seeking to open new food business respecting healthcare conditions. By using a simple and efficient clustering mechanism (kmeans), we showed an example of promoting a neighborhood (e.g. Chelsea) to be to most probably successful candidate for the business goal.

## Conclusion

Nowadays, machine learning is considered a core leveraging technology which can be used in every field where data exists and plays an important role. In this project, we applied a very simple unsupervised machine learning algorithm (i.e. Kmeans) in order to solve a business problem consisting of opening a new vegetarian restaurant in the most appropriate neighborhood of New York city/ Manhattan borough. We showed, using one use case that our methodology is successful in finding a perfect location for our planned business.