

Abstract	
Introduction	1
<i>Background</i>	1
<i>Problem</i>	1
<i>Stakeholders</i>	1
Methodology	
<i>Data Sources</i>	2
<i>NYU Spatial Data Repository</i>	2
<i>Foursquare 'Places API'</i>	2
<i>Data Retrieval</i>	
<i>Neighborhood Name & Location Data</i>	6
<i>Foursquare Food and Restaurant Related Venue Data</i>	7
<i>Exploratory Data Analysis</i>	8
<i>Data PreProcessing</i>	10
<i>Data Cleaning</i>	10
<i>OneHotEncoding Venue Categories</i>	10
<i>Feature Generation</i>	12
Results	
<i>Cluster Modeling</i>	13
<i>Cluster Visualization</i>	14
<i>Cluster Evaluation</i>	15
Conclusion	16
References	17
Appendix	18
<i>Clusters</i>	18

Abstract

Machine learning allows for the creation of computational models capable of identifying patterns in multidimensional datasets. I have tried to leverage venue data from Foursquare's 'Places API' and a machine learning algorithm called 'kmeans clustering' to identify 'New York City' neighborhoods of similarity towards food and restaurant business profiles'.

Introduction

Background

For the whole world the year 2007 to 2008 was known as the global financial crisis which unsurprisingly severely affected the restaurant industry in the United States. In mid 2008 A CNN Money report noted that the casual dining chains were taking a major hit as people turned to cheaper food alternatives.¹ Unsurprisingly, this state of economic unrest severely affected the restaurant industry. A number of restaurant businesses filed for bankruptcy and closed their doors. At the same time, the job market was just as bleak, with unemployment rates reaching up to 10% in the years following the Great Recession. After the economy bounced back it was reported increased US food sales year after year, with 2017 total retail and food services sales reaching 5.75 trillion in 2019.²

According to NRA (National Restaurants Association), as of 2017 there are over one million foods and restaurant service in the united states among these 26,697 were located in NYC. The restaurant industry shows in increase of 62% between 2000 and 2015.

Problem

The decision to open your own restaurant is an exciting one. You get to be your own boss and you do not have to worry about making money for anyone other than yourself. Being independent has its perks but it also has its costs-literally and figuratively. There are two ways that you can satisfy your dream, either you have to buy the existing restaurant or start a new restaurant. In either way one of the main challenging questions is which neighborhood is best for which type of food and restaurant service type. The project try to quantify food related services in NYC neighbor hoods and cluster them in their similarity characteristics so that it helps to make informed decision to start new business in food and restaurant service

Stakeholders

From the model, quantifying NYC neighborhoods similarities based on their food and restaurant service venues, developed the following stakeholders might be interested to got some informed decision making points.

- Business owners : Those who wants to open a new restaurants in NYC this model will help them to identify which neighborhood is more interested in whci type of food or restaurant service.
- Private and government service sector company – Some company might be interested in particular type of restaurant when they open their office in particular area for the sake of their employee.

Methodology

Data Sources

NYU Spatial Data Repository : ³

New York City Neighborhood Names 2014: This New York City Neighborhood Names point file was created as a guide to New York City's neighborhoods that appear on the web resource, "New York: A City of Neighborhoods." NYU spatial data .

Foursquare 'Places API' ⁴

Foursquare 'Places API' I will be using Foursquare's 'Places API' to acquire data related to 'venues' (as defined by Foursquare) categorized to be somehow associated with food and restaurants. It is important to note that foursquare defines a 'venue' as a place that one can go to, or checkin to, and that a 'venue' is not necessarily food or restaurant venue but can be any establishment such as a music or type of retail shop. Each Foursquare 'venue' is assigned a 'category' and each 'category' is associated with a particular 'categoryID'. The list of food and restaurant related categories included in this project are presented below.

Table : 1 List of food and restaurant related categories and their categoryid as presented in Foursqaure

Category	CategoryID
Afghan Restaurant	503288ae91d4c4b30a586d67
African Restaurant	4bf58dd8d48988d1c8941735
American Restaurant	4bf58dd8d48988d14e941735
Asian Restaurant	4bf58dd8d48988d142941735
Filipino Restaurant	4eb1bd1c3b7b55596b4a748f
Himalayan Restaurant	52e81612bcb5c57f1066b79fb
Hotpot Restaurant	52af0bd33cf9994f4e043bdd
Indonesian Restaurant	4deefc054765f83613cdba6f
Japanese Restaurant	4bf58dd8d48988d111941735
Korean Restaurant	4bf58dd8d48988d113941735
Malay Restaurant	4bf58dd8d48988d156941735
Mongolian Restaurant	4eb1d5724b900d56c88a45fe
Noodle House	4bf58dd8d48988d1d1941735
Satay Restaurant	56aa371be4b08b9a8d57350e
Thai Restaurant	4bf58dd8d48988d149941735
Tibetan Restaurant	52af39fb3cf9994f4e043be9
Australian Restaurant	4bf58dd8d48988d169941735
Austrian Restaurant	52e81612bcb5c57f1066b7a01
BBQ Joint	4bf58dd8d48988d1df931735
Bagel Shop	4bf58dd8d48988d179941735
Bakery	4bf58dd8d48988d16a941735

Belgian Restaurant	52e81612bcbc57f1066b7a02
Bistro	52e81612bcbc57f1066b79f1
Breakfast Spot	4bf58dd8d48988d143941735
Bubble Tea Shop	52e81612bcbc57f1066b7a0c
Buffet	52e81612bcbc57f1066b79f4
Burger Joint	4bf58dd8d48988d16c941735
Cafeteria	4bf58dd8d48988d128941735
Café	4bf58dd8d48988d16d941735
Cajun / Creole Restaurant	4bf58dd8d48988d17a941735
Caribbean Restaurant	4bf58dd8d48988d144941735
Caucasian Restaurant	5293a7d53cf9994f4e043a45
Coffee Shop	4bf58dd8d48988d1e0931735
Comfort Food Restaurant	52e81612bcbc57f1066b7a00
Creperie	52e81612bcbc57f1066b79f2
Czech Restaurant	52f2ae52bcbc57f1066b8b81
Deli / Bodega	4bf58dd8d48988d146941735
Dessert Shop	4bf58dd8d48988d1d0941735
Diner	4bf58dd8d48988d147941735
Donut Shop	4bf58dd8d48988d148941735
Dumpling Restaurant	4bf58dd8d48988d108941735
Dutch Restaurant	5744ccdf4b0c0459246b4d0
Eastern European Restaurant	4bf58dd8d48988d109941735
English Restaurant	52e81612bcbc57f1066b7a05
Falafel Restaurant	4bf58dd8d48988d10b941735
Fast Food Restaurant	4bf58dd8d48988d16e941735
Fish & Chips Shop	4edd64a0c7ddd24ca188df1a
Fondue Restaurant	52e81612bcbc57f1066b7a09
Food Court	4bf58dd8d48988d120951735
Food Stand	56aa371be4b08b9a8d57350b
Food Truck	4bf58dd8d48988d1cb941735
French Restaurant	4bf58dd8d48988d10c941735
Fried Chicken Joint	4d4ae6fc7a7b7dea34424761
Friterie	55d25775498e9f6a0816a37a
Gastropub	4bf58dd8d48988d155941735
German Restaurant	4bf58dd8d48988d10d941735
Gluten-free Restaurant	4c2cd86ed066bed06c3c5209
Greek Restaurant	4bf58dd8d48988d10e941735
Halal Restaurant	52e81612bcbc57f1066b79ff
Hawaiian Restaurant	52e81612bcbc57f1066b79fe
Hot Dog Joint	4bf58dd8d48988d16f941735
Hungarian Restaurant	52e81612bcbc57f1066b79fa
Indian Restaurant	4bf58dd8d48988d10f941735
Irish Pub	52e81612bcbc57f1066b7a06

Italian Restaurant	4bf58dd8d48988d110941735
Jewish Restaurant	52e81612bcbc57f1066b79fd
Juice Bar	4bf58dd8d48988d112941735
Kebab Restaurant	5283c7b4e4b094cb91ec88d7
Latin American Restaurant	4bf58dd8d48988d1be941735
Mac & Cheese Joint	4bf58dd8d48988d1bf941735
Mediterranean Restaurant	4bf58dd8d48988d1c0941735
Mexican Restaurant	4bf58dd8d48988d1c1941735
Middle Eastern Restaurant	4bf58dd8d48988d115941735
Modern European Restaurant	52e81612bcbc57f1066b79f9
Molecular Gastronomy Restaurant	4bf58dd8d48988d1c2941735
Pakistani Restaurant	52e81612bcbc57f1066b79f8
Pet Café	56aa371be4b08b9a8d573508
Pizza Place	4bf58dd8d48988d1ca941735
Polish Restaurant	52e81612bcbc57f1066b7a04
Portuguese Restaurant	4def73e84765ae376e57713a
Poutine Place	56aa371be4b08b9a8d5734c7
Restaurant	4bf58dd8d48988d1c4941735
Russian Restaurant	5293a7563cf9994f4e043a44
Salad Place	4bf58dd8d48988d1bd941735
Sandwich Place	4bf58dd8d48988d1c5941735
Scandinavian Restaurant	4bf58dd8d48988d1c6941735
Scottish Restaurant	5744ccdde4b0c0459246b4a3
Seafood Restaurant	4bf58dd8d48988d1ce941735
Slovak Restaurant	56aa371be4b08b9a8d57355a
Snack Place	4bf58dd8d48988d1c7941735
Soup Place	4bf58dd8d48988d1dd931735
Southern / Soul Food Restaurant	4bf58dd8d48988d14f941735
Spanish Restaurant	4bf58dd8d48988d150941735
Sri Lankan Restaurant	5413605de4b0ae91d18581a9
Steakhouse	4bf58dd8d48988d1cc941735
Swiss Restaurant	4bf58dd8d48988d158941735
Tea Room	4bf58dd8d48988d1dc931735
Theme Restaurant	56aa371be4b08b9a8d573538
Truck Stop	57558b36e4b065eceed306dd
Turkish Restaurant	4f04af1f2fb6e1c99f3db0bb
Ukrainian Restaurant	52e928d0bcbc57f1066b7e96
Vegetarian / Vegan Restaurant	4bf58dd8d48988d1d3941735
Wings Joint	4bf58dd8d48988d14c941735

Data Retrieval

Neighborhood Name & Location Data The '2014 New York City neighborhood Names' dataset hosted by NYU's Spatial Data Repository was first downloaded as a JSON file and imported into a Jupyter Notebook to merge with Foursquare Food and Restaurant Related Venue Data

```
with open('NYC data set.json') as json_data:
    newyork_data = json.load(json_data)
nyc_neighborhood_data = newyork_data['features']
nyc_neighborhood_data[5]

{'type': 'Feature',
 'id': 'nyu_2451_34572.6',
 'geometry': {'type': 'Point',
 'coordinates': [-73.90281798724604, 40.88168737120521]},
 'geometry_name': 'geom',
 'properties': {'name': 'Kingsbridge',
 'stacked': 1,
 'annoline1': 'Kingsbridge',
 'annoline2': None,
 'annoline3': None,
 'annoangle': 0.0,
 'borough': 'Bronx',
 'bbox': [-73.90281798724604,
 40.88168737120521,
 -73.90281798724604,
 40.88168737120521]}}
```

Neighborhood Name & Location Data

The 'Borough', 'Neighborhood', 'Latitude', and 'Longitude' values associated with each neighborhood were then converted from JSON to a Pandas DataFrame that serves as the foundation of the analysis.

```
column_names = ['Borough', 'Neighborhood', 'Latitude', 'Longitude']

nyc_neighborhoods = pd.DataFrame(columns=column_names)

for data in nyc_neighborhood_data:
    borough = nyc_neighborhood_data = data['properties']['borough']
    neighborhood_name = data['properties']['name']

    neighborhood_latlon = data['geometry']['coordinates']
    neighborhood_lat = neighborhood_latlon[1]
    neighborhood_lon = neighborhood_latlon[0]

    nyc_neighborhoods = nyc_neighborhoods.append({'Borough': borough,
                                                    'Neighborhood': neighborhood_name,
                                                    'Latitude': neighborhood_lat,
                                                    'Longitude': neighborhood_lon}, ignore_index=True)

nyc_neighborhoods.head(5)
```

Foursquare Food and Restaurant Related Venue Data

Food and restaurant related venues associated with each New York City neighborhood was created by recursively sending ‘get’ requests to the predefined endpoint, making sure the results are specific to venues with food and restaurant related ‘category IDs’. For each neighborhood, we can include all of the selected category IDs in a single ‘get’ request by passing them as comma separated values. Shown below is a function that creates the required url and an example:

Function To Retrieve All Food-Related Venues Per Neighborhood

```
] M radius = 100
def getNearbyFoodVenues(neighborhoods, latitudes, longitudes, radius=100):
    endpoint = 'https://api.foursquare.com/v2/venues/search?'
    venues_list = []

    for hood_name, lat, lng in zip(neighborhoods, latitudes, longitudes):
        url = createURL(endpoint, CLIENT_ID, CLIENT_SECRET, VERSION, lat, lng, radius, categoryId, limit)
        results = requests.get(url).json()['response']['venues']

        for item in results:
            venue_name = item['name']
            venue_category = item['categories'][0]['name']
            venue_lat = item['location']['lat']
            venue_lng = item['location']['lng']
            try:
                venue_city = item['location']['city']
            except:
                venue_city = 'N/A'
            venue_state = item['location']['state']
            venues_list.append([(hood_name,
                                lat,
                                lng,
                                venue_name,
                                venue_category,
                                venue_lat,
                                venue_lng,
                                venue_city,
                                venue_state
                                )])

    nearby_venues = pd.DataFrame([item for venue_list in venues_list for item in venue_list])
    nearby_venues.columns = ['Neighborhood',
                             'Neighborhood Latitude',
                             'Neighborhood Longitude',
                             'Venue Name',
                             'Venue Category',
                             'Venue Latitude',
                             'Venue Longitude',
                             'Venue City',
                             'Venue State'
                             ]
    return nearby_venues

venues_data = getNearbyFoodVenues(nyc_neighborhoods['Neighborhood'],
                                   nyc_neighborhoods['Latitude'],
                                   nyc_neighborhoods['Longitude'],
                                   radius)
```

Calculating the number of venues We Retrieved?

```
print(venues_data.shape)
venues_data.head(5)
```

(1503, 9)

[9]:

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue Name	Venue Category	Venue Latitude	Venue Longitude	Venue City	Venue State
0	Kingsbridge	40.881687	-73.902818	Silhouette Restaurant & Lounge	Latin American Restaurant	40.880706	-73.902887	Bronx	NY
1	Kingsbridge	40.881687	-73.902818	Ocean Dip Seafood	Seafood Restaurant	40.880771	-73.903375	Bronx	NY
2	Kingsbridge	40.881687	-73.902818	Vickies Pizza	Pizza Place	40.881003	-73.903336	Bronx	NY
3	Woodlawn	40.898273	-73.867315	Rambling House	Pub	40.898439	-73.867197	Bronx	NY
4	Woodlawn	40.898273	-73.867315	Sean's Quality Deli	Deli / Bodega	40.897595	-73.867151	Bronx	NY

Our dataframe has 1503 venues pulled from foursquare.

Exploratory Data Analysis

The dataframe has shown that the venues are belonged into three venue state namely NJ, NY and New York with frequency of 2, 1463 and 38 respectively.

Check the states where the venues belongs in?

```
venues_data.groupby('Venue State')['Venue State'].count()
```

[12]: Venue State
NJ 2
NY 1463
New York 38

The dataframe has shown also there are 121 uniuqe venue categories with 187 of maximum number of venue in it and the minimum of 1.


```
n_unique = len(venues_data['Venue Category'].unique())
print(f'There are {n_unique} unique venue categories in this dataframe')
venues_data.groupby('Venue Category')['Venue Category'].count().sort_values(ascending=False)
```

There are 121 unique venue categories in this dataframe

Venue Category	
Deli / Bodega	187
Pizza Place	100
Food Truck	86
Chinese Restaurant	81
Coffee Shop	77
Bakery	52
Café	49
Italian Restaurant	48
American Restaurant	44
Mexican Restaurant	40
Ice Cream Shop	34
Sandwich Place	32
Restaurant	29
Juice Bar	27
Caribbean Restaurant	27
Diner	26
Bagel Shop	25
Korean Restaurant	22
Burger Joint	21
Dessert Shop	20
Sushi Restaurant	19
Thai Restaurant	18
Donut Shop	17
Fried Chicken Joint	17
Japanese Restaurant	16
Vegetarian / Vegan Restaurant	16
Fast Food Restaurant	16
Indian Restaurant	16
Greek Restaurant	16
Asian Restaurant	14
...	...
Bike Shop	1
Bistro	1
Yoga Studio	1
Comfort Food Restaurant	1
Newsstand	1
..	.

There are 38 venues which has no venue cities not mentioned. These records has any meaningful impact for our model analysis so that we are going to remove from our dataframe.

Are there any 'null' values in the dataset?

```
venues_data.isnull().values.any()
```

6]: True

```
null_counts= venues_data.isnull()
print(null_counts)
```

```
null_counts = null_counts.sum(axis=0)
null_counts
```

```
Neighborhood      0
Neighborhood Latitude  0
Neighborhood Longitude  0
Venue Name        0
Venue Category    0
Venue Latitude    0
Venue Longitude   0
Venue City        38
Venue State       0
dtype: int64
```

How many unique venues are there?

```
[17]: n_unique_venues = len(venues_data['Venue Name'].unique())
      print(f'There are {n_unique_venues} unique venues in the venues dataset')

      There are 1409 unique venues in the venues dataset
```

The dataframe has 1409 unique venues.

Data Preprocessing

Data Cleaning

- We have removed all venue records with venue city name different from “NY” or “New York” from the data frame.*
- We have dropped all NaN values from venue city*

The dataframe shows there are 38 Nan records. So we have remove them as it has no impact for our model.

```
null_counts = venues_data.isnull()
null_counts = null_counts.sum(axis=0)
null_counts
```

```
Neighborhood      0
Neighborhood Latitude  0
Neighborhood Longitude  0
Venue Name        0
Venue Category    0
Venue Latitude    0
Venue Longitude   0
Venue City        38
Venue State       0
dtype: int64
```

I have used the dropna function to drop nan values

```
venue_data = venues_data.dropna(how='any')
venue_data
```

```

Neighborhood      0
Neighborhood Latitude  0
Neighborhood Longitude  0
Venue Name        0
Venue Category    0
Venue Latitude    0
Venue Longitude   0
Venue City       38
Venue State      0
dtype: int64

```

- c) The list of venue category predefined for make our analysis easy. I have also deleted all other category other than predefined venue categories which are related with food and restaurant services.

Remove entries that are not associated with food or resutrant categories

```

food_related_categories = ['Afghan Restaurant', 'African Restaurant', 'American Restaurant', 'Asian Restaurant', 'Filipino Restaurant', 'Himalayan Restaurant', 'Hotpot Restaurant', 'Indonesian Restaurant', 'Japanese Restaurant', 'Korean Restaurant', 'Malay Restaurant', 'Mongolian Restaurant', 'Noodle House', 'Satay Restaurant', 'Thai Restaurant', 'Tibetan Restaurant', 'Austrian Restaurant', 'BBQ Joint', 'Bagel Shop', 'Bakery', 'Belgian Restaurant', 'Bistro', 'Breakfast Spot', 'Bubble Tea Shop', 'Buffet', 'Burger Joint', 'Cafeteria', 'Café', 'Cajun / Creole Restaurant', 'Caribbean Restaurant', 'Caucasian Restaurant', 'Coffee Shop', 'Comfort Food Restaurant', 'Creperie', 'Czech Restaurant', 'Deli / Bodega', 'Dessert Shop', 'Diner', 'Donut Shop', 'Dumpling Restaurant', 'Dutch Restaurant', 'Eastern European Restaurant', 'English Restaurant', 'Falafel Restaurant', 'Fast Food Restaurant', 'Fish & Chips Shop', 'Fondue Restaurant', 'Food Court', 'Food Stand', 'Food Truck', 'French Restaurant', 'Fried Chicken Joint', 'Friterie', 'Gastropub', 'German Restaurant', 'Gluten-free Restaurant', 'Greek Restaurant', 'Halal Restaurant', 'Hawaiian Restaurant', 'Hot Dog Joint', 'Hungarian Restaurant', 'Indian Restaurant', 'Irish Pub', 'Italian Restaurant', 'Jewish Restaurant', 'Juice Bar', 'Kebab Restaurant', 'Latin American Restaurant', 'Mac & Cheese Joint', 'Mediterranean Restaurant', 'Mexican Restaurant', 'Middle Eastern Restaurant', 'Modern European Restaurant', 'Molecular Gastronomy Restaurant', 'Pakistani Restaurant', 'Pet Café', 'Pizza Place', 'Polish Restaurant', 'Portuguese Restaurant', 'Poutine Place', 'Restaurant', 'Russian Restaurant', 'Salad Place', 'Sandwich Place', 'Scandinavian Restaurant', 'Scottish Restaurant', 'Seafood Restaurant', 'Slovak Restaurant', 'Snack Place', 'Soup Place', 'Southern / Soul Food Restaurant', 'Spanish Restaurant', 'Sri Lankan Restaurant', 'Steakhouse', 'Swiss Restaurant', 'Tea Room', 'Theme Restaurant', 'Truck Stop', 'Turkish Restaurant', 'Ukrainian Restaurant', 'Vegetarian / Vegan Restaurant', 'Wings Joint']
ny_food_venues = ny_venue_data_with_city[ny_venue_data_with_city['Venue Category'].isin(food_related_categories)]
delta = ny_venue_data_with_city.shape[0] - ny_food_venues.shape[0]
print(f'{delta} entries were removed based on "Venue Category" not being related to food')
print(ny_food_venues['Venue Category'].unique())
ny_food_venues.head(5)

```

- d) The image below shows the total number of entries and number of unique entries in the ny_foodc_venues dataframe. As previously mentioned, some venues are assigned to multiple neighborhoods because the venue is within 500 meters of the neighborhood's centroid location.

How many food venue entries are left?

```

n_entries = ny_food_venues.shape[0]
print(f'There are {n_entries} entries in the ny_food_venues dataframe')

```

There are 1271 entries in the ny_food_venues dataframe

How many unique food venues are left?

```

n_unique_entries = len(ny_food_venues['Venue Name'].unique())
print(f'There are {n_unique_entries} unique entries in the ny_food_venues dataframe')

```

There are 1188 unique entries in the ny_food_venues dataframe

OneHotEncoding Venue Categories

In order to use Foursquare's category values to find similar neighborhoods based on food venues, a onehotencoding representation of each entry was created using Pandas' 'get_dummies' function. The result was a dataframe of New York City food elated venues

where entry venue category is represented by a value of 1 in the column of matching venue category:

One-Hot-Encode Venue Categories

```
ny_venue_category_onehot = pd.get_dummies(ny_food_venues[['Venue Category']], prefix="", prefix_sep="")

ny_venue_category_onehot['Neighborhood'] = ny_food_venues['Neighborhood']

fixed_columns = [ny_venue_category_onehot.columns[-1]] + list(ny_venue_category_onehot.columns[:-1])
ny_venue_category_onehot = ny_venue_category_onehot[fixed_columns]

print(ny_venue_category_onehot.shape)
ny_venue_category_onehot.head()

(1271, 74)
```

7]:

	Neighborhood	Afghan Restaurant	African Restaurant	American Restaurant	Asian Restaurant	BBQ Joint	Bagel Shop	Bakery	Bistro	Breakfast Spot	...	Snack Place	Southern / Soul Food Restaurant	Spanish Restaurant	Sri Lankan Restaurant
0	Kingsbridge	0	0	0	0	0	0	0	0	0	...	0	0	0	0
1	Kingsbridge	0	0	0	0	0	0	0	0	0	...	0	0	0	0
2	Kingsbridge	0	0	0	0	0	0	0	0	0	...	0	0	0	0
4	Woodlawn	0	0	0	0	0	0	0	0	0	...	0	0	0	0
5	Woodlawn	0	0	0	0	0	0	0	0	0	...	0	0	0	0

5 rows x 74 columns

Feature Generation

After encoding the dataset of food and restaurant related in New York city then used to quantify to assess the profile of each neighborhood. For each venue category, the percent distribution of venues across each neighborhood was calculated. This information would then be used to fit a KMeans clustering algorithm to the data in an effort to determine neighborhoods of similar food venue profile.

For each venue category, determine the percentage of entities in each neighborhood

```
venue_mean = pd.DataFrame()
for category, total in venue_totals.items():
    venue_mean[category] = venue_counts[category].apply(lambda x: x / total)
venue_mean = venue_mean.reindex(sorted(venue_mean.columns), axis=1).reset_index()
venue_mean.head(5)
```

8]:

	Neighborhood	Afghan Restaurant	African Restaurant	American Restaurant	Asian Restaurant	Filipino Restaurant	Himalayan Restaurant	Hotpot Restaurant
0	Allerton	0.0	0.0	0.0	0.000000	0.0	0.0	0.0
1	Arlington	0.0	0.0	0.0	0.000000	0.0	0.0	0.0
2	Arrochar	0.0	0.0	0.0	0.000000	0.0	0.0	0.0
3	Astoria	0.0	0.0	0.0	0.000000	0.0	0.0	0.0
4	Bath Beach	0.0	0.0	0.0	0.071429	0.0	0.0	0.0

Creating a dataframe that shows the top 5 venue categories each neighborhood is known for

```

def return_top_venue_categories(row, num_top_venues):
    row_categories = row.iloc[1:]
    row_categories_sorted = row_categories.sort_values(ascending=False)

    return row_categories_sorted.index.values[0:num_top_venues]

num_top_venues = 5

indicators = ['st', 'nd', 'rd']

# create columns according to number of top venues
columns = ['Neighborhood']
for ind in np.arange(num_top_venues):
    try:
        columns.append('{}{} Top Venue Category'.format(ind+1, indicators[ind]))
    except:
        columns.append('{}th Top Venue Category'.format(ind+1))

neighborhoods_top_venue_categories = pd.DataFrame(columns=columns)
neighborhoods_top_venue_categories['Neighborhood'] = venue_mean['Neighborhood']

for ind in np.arange(venue_mean.shape[0]):
    neighborhoods_top_venue_categories.iloc[ind, 1:] = return_top_venue_categories(venue_mean.iloc[ind, :], num_top_venues)

neighborhoods_top_venue_categories.head(5)

```

```

9]:

```

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category
0	Allerton	Hotpot Restaurant	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	American Restaurant
1	Arlington	Hotpot Restaurant	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	American Restaurant
2	Arrochar	Hotpot Restaurant	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	American Restaurant
3	Astoria	Hotpot Restaurant	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	American Restaurant
4	Bath Beach	Asian Restaurant	Hotpot Restaurant	Himalayan Restaurant	Filipino Restaurant	American Restaurant

Results

Cluster Modeling

Scikitlearn's KMeans clustering was used to determine similar neighborhoods based on food venue percentage. The image below shows the data being scaled and the KMeans model being created:

Create a KMeans Model To Cluster Neighborhoods

```

kclusters = 15

venue_grouped_clustering = venue_mean.drop('Neighborhood', 1)
kmeans = KMeans(n_clusters=kclusters, random_state=0).fit(venue_grouped_clustering)

kmeans.labels_

```

```

0]: array([ 0,  0,  0,  0,  7,  8,  0,  8,  0,  0,  0,  0,  0,  0,  0,  0,  0,
           0,  0,  0,  0,  7,  0,  0,  0,  8,  0,  0,  0,  8,  0,  0,  0,  7,
           0,  0,  0,  0,  0,  0,  0,  0,  8,  0,  0,  0, 12,  0,  0,  0,  0,
           5,  4,  0,  0,  0,  0,  0,  0,  0,  8,  8,  0,  7,  0,  0,  0,  0,
           0, 11,  0,  0,  0, 12,  8,  0,  0,  8,  8,  0,  7,  0, 10,  0,  0,
           6,  0, 14,  8,  0,  0,  0,  0,  8,  1,  0,  2,  8,  0,  0,  0,
           0,  0,  0,  0,  0,  0,  9, 11,  0,  0,  0,  0,  0,  0,  8,  0,  5,
           3,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  6,
           0,  0,  8,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  8,  0,  0,
           0,  8,  0,  8,  0,  0,  8,  0,  0,  0,  0,  8,  0,  0,  0,  0,  4,
           8,  0,  0, 11,  0,  0,  0,  0,  0,  0, 13,  0])

```

A new dataframe was created by merging neighborhood location data with cluster labels and top venue categories.

Creating a new dataframe merging neighborhood location data, top venue category by neighborhood, and cluster labels

```
: | #neighborhoods_top_venue_categories.insert(1, 'Cluster Labels', kmeans.Labels_)
ny_neighborhood_food_profile = nyc_neighborhoods.drop(columns=['Borough'])
ny_neighborhood_food_profile = ny_neighborhood_food_profile.join(neighborhoods_top_venue_categories.set_index('Neighborhood'))
ny_neighborhood_food_profile.head()
```

Cluster Visualization

I have used folium visualization package to visualize clustered similar food and restaurant neighborhoods in New York city.

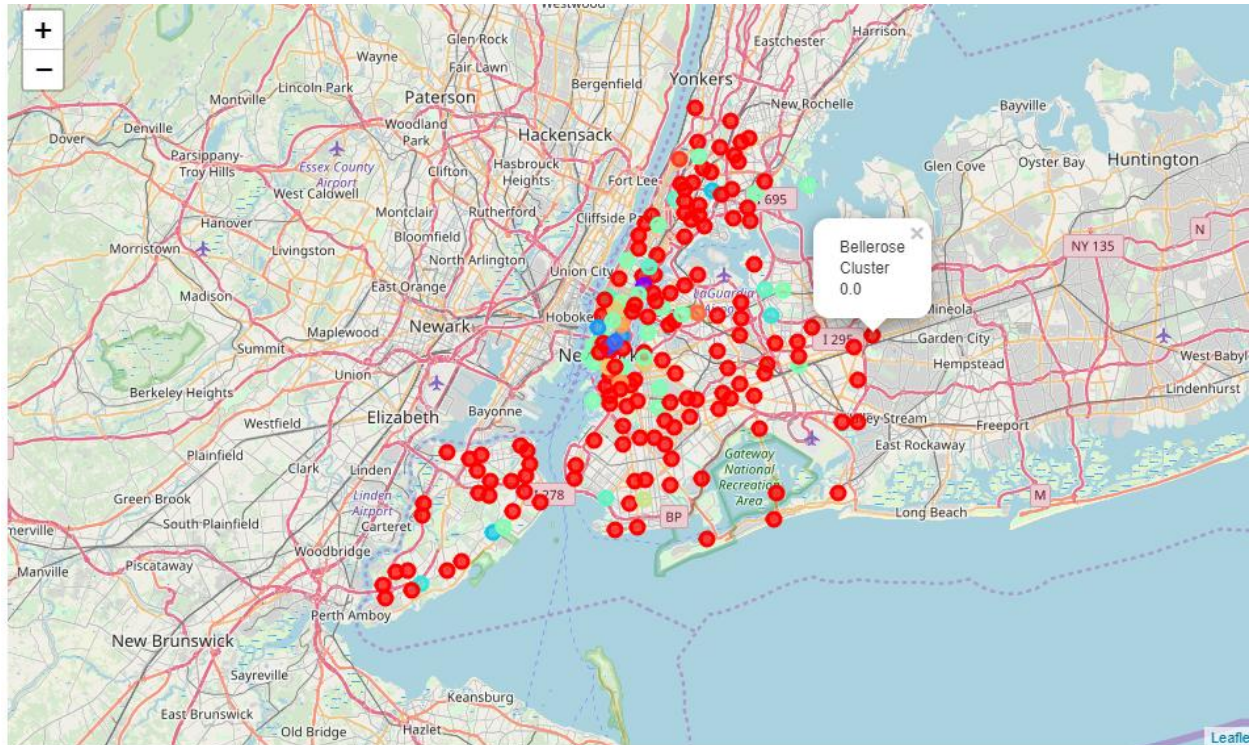
Creating A Visualization of The Clusters

```
In [90]: | # create map
import matplotlib.cm as cm
import matplotlib.colors as colors
latitude = 40.730610
longitude = -73.935242
map_clusters = folium.Map(location=[latitude, longitude], zoom_start=10.45)

# set color scheme for the clusters
x = np.arange(kclusters)
ys = [i + x + (i*x)**2 for i in range(kclusters)]
colors_array = cm.rainbow(np.linspace(0, 1, len(ys)))
rainbow = [colors.rgb2hex(i) for i in colors_array]

# add markers to the map
markers_colors = []
for lat, lon, poi, cluster in zip(ny_neighborhood_food_profile['Latitude'], ny_neighborhood_food_profile['Longitude'], ny_neighborhood_food_profile['Top Venue Category'], ny_neighborhood_food_profile['Cluster Labels']):
    label = folium.Popup(str(poi) + ' Cluster ' + str(cluster), parse_html=True)
    folium.CircleMarker(
        [lat, lon],
        radius=5,
        popup=label,
        color=rainbow[int(cluster)-1],
        fill=True,
        fill_color=rainbow[int(cluster)-1],
        fill_opacity=0.7).add_to(map_clusters)

map_clusters
```



Cluster Evaluation

I have the following coded which iterates through and prints the results of each cluster:

```
for cluster in range(0, kclusters):
    print(f'Cluster {cluster}:')
    print(ny_neighborhood_food_profile.loc[ny_neighborhood_food_profile['Cluster Labels'] == cluster, ny_neighborhood_food_profile.columns])
```

The cluster detail list is presented at the appendix section. Let's try see some interesting results showed up from the kmeans clustering algorithm.

The majority of food and restaurant related services are accumulated in the first cluster and ninth cluster.

In the first cluster Hotspot restaurants are more abundant with Himalayan Restaurant, Filipino Restaurant, Asian Restaurant and American Restaurant following consecutively.

From cluster 2 upto cluster 7, it seems that there is no much food and restaurant venues. Each of them have two or one venues.

From cluster 10 to 14 also lower number of food and restaurant related venues.

Conclusion

Machine learning algorithm can applied to different multidimensional dataset to find meaningful pattern and similarities among data. Cluster of food and restaurant related service in given city can generated with high level geo spatial data with additional parameters. But, in this project for leaning purpose we have limited radius as 500 and amount of data limit as 25. The result has shown clearly which areas are more prone for food and restaurant services and what type of restaurants' are more popular.

These findings can used start up and baseline for business owner who wants to start new restaurant business to conduct further investigation.

References

¹https://money.cnn.com/2008/07/29/news/companies/restaurant_industry/index.htm?postversion=2008072919

²<https://data.bls.gov/timeseries/LNS14000000>

³ 2014 New York City Neighborhood Names NYU Spatial Data Repository

⁴ 'Places API' Documentation Foursquare

Appendix

Clusters

Cluster 0:

	Neighborhood	1st Top Venue	Category	2nd Top Venue	Category	\
5	Kingsbridge	Hotpot	Restaurant	Himalayan	Restaurant	
7	Woodlawn	Hotpot	Restaurant	Himalayan	Restaurant	
8	Norwood	Hotpot	Restaurant	Himalayan	Restaurant	
9	Williamsbridge	Hotpot	Restaurant	Himalayan	Restaurant	
15	Morris Heights	Hotpot	Restaurant	Himalayan	Restaurant	
16	Fordham	Hotpot	Restaurant	Himalayan	Restaurant	
18	West Farms	Hotpot	Restaurant	Himalayan	Restaurant	
20	Melrose	Hotpot	Restaurant	Himalayan	Restaurant	
21	Mott Haven	Hotpot	Restaurant	Himalayan	Restaurant	
23	Longwood	Hotpot	Restaurant	Himalayan	Restaurant	
25	Morrisania	Hotpot	Restaurant	Himalayan	Restaurant	
26	Soundview	Hotpot	Restaurant	Himalayan	Restaurant	
32	Van Nest	Hotpot	Restaurant	Himalayan	Restaurant	
34	Belmont	Hotpot	Restaurant	Himalayan	Restaurant	
36	North Riverdale	Hotpot	Restaurant	Himalayan	Restaurant	
37	Pelham Bay	Hotpot	Restaurant	Himalayan	Restaurant	
40	Castle Hill	Hotpot	Restaurant	Himalayan	Restaurant	
41	Olinville	Hotpot	Restaurant	Himalayan	Restaurant	
43	Concourse	Hotpot	Restaurant	Himalayan	Restaurant	
44	Unionport	Hotpot	Restaurant	Himalayan	Restaurant	
45	Edenwald	Hotpot	Restaurant	Himalayan	Restaurant	
46	Bay Ridge	Hotpot	Restaurant	Himalayan	Restaurant	
48	Sunset Park	Hotpot	Restaurant	Himalayan	Restaurant	
50	Gravesend	Hotpot	Restaurant	Himalayan	Restaurant	
51	Brighton Beach	Hotpot	Restaurant	Himalayan	Restaurant	
53	Manhattan Terrace	Hotpot	Restaurant	Himalayan	Restaurant	
56	East Flatbush	Hotpot	Restaurant	Himalayan	Restaurant	
57	Kensington	Hotpot	Restaurant	Himalayan	Restaurant	
58	Windsor Terrace	Hotpot	Restaurant	Himalayan	Restaurant	
59	Prospect Heights	Hotpot	Restaurant	Himalayan	Restaurant	
..	
231	Dongan Hills	Hotpot	Restaurant	Himalayan	Restaurant	
239	Charleston	Hotpot	Restaurant	Himalayan	Restaurant	
244	Chelsea	Hotpot	Restaurant	Himalayan	Restaurant	
249	Civic Center	Hotpot	Restaurant	Himalayan	Restaurant	
254	Concord	Hotpot	Restaurant	Himalayan	Restaurant	
258	Elm Park	Hotpot	Restaurant	Himalayan	Restaurant	
263	Jamaica Hills	Hotpot	Restaurant	Himalayan	Restaurant	
267	Claremont Village	Hotpot	Restaurant	Himalayan	Restaurant	
268	Concourse Village	Hotpot	Restaurant	Himalayan	Restaurant	
269	Mount Eden	Hotpot	Restaurant	Himalayan	Restaurant	
270	Mount Hope	Hotpot	Restaurant	Himalayan	Restaurant	
272	Hunters Point	Hotpot	Restaurant	Himalayan	Restaurant	
273	Turtle Bay	Hotpot	Restaurant	Himalayan	Restaurant	
274	Tudor City	Hotpot	Restaurant	Himalayan	Restaurant	
278	Blissville	Hotpot	Restaurant	Himalayan	Restaurant	
281	Weeksville	Hotpot	Restaurant	Himalayan	Restaurant	
282	Broadway Junction	Hotpot	Restaurant	Himalayan	Restaurant	
283	Dumbo	Hotpot	Restaurant	Himalayan	Restaurant	
284	Manor Heights	Hotpot	Restaurant	Himalayan	Restaurant	
285	Willowbrook	Hotpot	Restaurant	Himalayan	Restaurant	
286	Sandy Ground	Hotpot	Restaurant	Himalayan	Restaurant	
288	Roxbury	Hotpot	Restaurant	Himalayan	Restaurant	
290	Middle Village	Hotpot	Restaurant	Himalayan	Restaurant	
291	Prince's Bay	Hotpot	Restaurant	Himalayan	Restaurant	
293	Richmond Valley	Hotpot	Restaurant	Himalayan	Restaurant	
298	Allerton	Hotpot	Restaurant	Himalayan	Restaurant	
300	Erasmus	Hotpot	Restaurant	Himalayan	Restaurant	
301	Hudson Yards	Hotpot	Restaurant	Himalayan	Restaurant	
304	Queensbridge	Hotpot	Restaurant	Himalayan	Restaurant	
305	Fox Hills	Hotpot	Restaurant	Himalayan	Restaurant	

	3rd Top Venue	Category	4th Top Venue	Category	5th Top Venue	Category
5	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
7	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
8	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
9	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
15	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
16	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
18	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
20	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
21	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
23	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
25	Filipino	Restaurant	Asian	Restaurant	American	Restaurant
26	Filipino	Restaurant	Asian	Restaurant	American	Restaurant

32	Filipino Restaurant	Asian Restaurant	American Restaurant
34	Filipino Restaurant	Asian Restaurant	American Restaurant
36	Filipino Restaurant	Asian Restaurant	American Restaurant
37	Filipino Restaurant	Asian Restaurant	American Restaurant
40	Filipino Restaurant	Asian Restaurant	American Restaurant
41	Filipino Restaurant	Asian Restaurant	American Restaurant
43	Filipino Restaurant	Asian Restaurant	American Restaurant
44	Filipino Restaurant	Asian Restaurant	American Restaurant
45	Filipino Restaurant	Asian Restaurant	American Restaurant
46	Filipino Restaurant	Asian Restaurant	American Restaurant
48	Filipino Restaurant	Asian Restaurant	American Restaurant
50	Filipino Restaurant	Asian Restaurant	American Restaurant
51	Filipino Restaurant	Asian Restaurant	American Restaurant
53	Filipino Restaurant	Asian Restaurant	American Restaurant
56	Filipino Restaurant	Asian Restaurant	American Restaurant
57	Filipino Restaurant	Asian Restaurant	American Restaurant
58	Filipino Restaurant	Asian Restaurant	American Restaurant
59	Filipino Restaurant	Asian Restaurant	American Restaurant
..
231	Filipino Restaurant	Asian Restaurant	American Restaurant
239	Filipino Restaurant	Asian Restaurant	American Restaurant
244	Filipino Restaurant	Asian Restaurant	American Restaurant
249	Filipino Restaurant	Asian Restaurant	American Restaurant
254	Filipino Restaurant	Asian Restaurant	American Restaurant
258	Filipino Restaurant	Asian Restaurant	American Restaurant
263	Filipino Restaurant	Asian Restaurant	American Restaurant
267	Filipino Restaurant	Asian Restaurant	American Restaurant
268	Filipino Restaurant	Asian Restaurant	American Restaurant
269	Filipino Restaurant	Asian Restaurant	American Restaurant
270	Filipino Restaurant	Asian Restaurant	American Restaurant
272	Filipino Restaurant	Asian Restaurant	American Restaurant
273	Filipino Restaurant	Asian Restaurant	American Restaurant
274	Filipino Restaurant	Asian Restaurant	American Restaurant
278	Filipino Restaurant	Asian Restaurant	American Restaurant
281	Filipino Restaurant	Asian Restaurant	American Restaurant
282	Filipino Restaurant	Asian Restaurant	American Restaurant
283	Filipino Restaurant	Asian Restaurant	American Restaurant
284	Filipino Restaurant	Asian Restaurant	American Restaurant
285	Filipino Restaurant	Asian Restaurant	American Restaurant
286	Filipino Restaurant	Asian Restaurant	American Restaurant
288	Filipino Restaurant	Asian Restaurant	American Restaurant
290	Filipino Restaurant	Asian Restaurant	American Restaurant
291	Filipino Restaurant	Asian Restaurant	American Restaurant
293	Filipino Restaurant	Asian Restaurant	American Restaurant
298	Filipino Restaurant	Asian Restaurant	American Restaurant
300	Filipino Restaurant	Asian Restaurant	American Restaurant
301	Filipino Restaurant	Asian Restaurant	American Restaurant
304	Filipino Restaurant	Asian Restaurant	American Restaurant
305	Filipino Restaurant	Asian Restaurant	American Restaurant

[139 rows x 6 columns]

Cluster 1:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
109	Lenox Hill	Afghan Restaurant	American Restaurant	
	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category	
109	Hotpot Restaurant	Himalayan Restaurant	Filipino Restaurant	

Cluster 2:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
121	Little Italy	Hotpot Restaurant	Himalayan Restaurant	
	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category	
121	Filipino Restaurant	Asian Restaurant	American Restaurant	

Cluster 3:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
248	Noho	Himalayan Restaurant	American Restaurant	
	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category	
248	Hotpot Restaurant	Filipino Restaurant	Asian Restaurant	

Cluster 4:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
118	East Village	Filipino Restaurant	American Restaurant	
123	West Village	Filipino Restaurant	American Restaurant	
	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category	
118	Hotpot Restaurant	Himalayan Restaurant	Asian Restaurant	
123	Hotpot Restaurant	Himalayan Restaurant	Asian Restaurant	

Cluster 5:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
17	East Tremont	African Restaurant	Hotpot Restaurant	
211	New Dorp	African Restaurant	Hotpot Restaurant	
	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category	
17	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
211	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	

Cluster 6:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
184	Queensboro Hill	Asian Restaurant	Hotpot Restaurant	
236	Huguenot	Asian Restaurant	Hotpot Restaurant	
	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category	
184	Himalayan Restaurant	Filipino Restaurant	American Restaurant	
236	Himalayan Restaurant	Filipino Restaurant	American Restaurant	

Cluster 7:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
19	High Bridge	Asian Restaurant	Hotpot Restaurant	
79	Bath Beach	Asian Restaurant	Hotpot Restaurant	
113	Clinton	Asian Restaurant	Hotpot Restaurant	
138	Flushing	Asian Restaurant	Hotpot Restaurant	
247	Carnegie Hill	Asian Restaurant	Hotpot Restaurant	
	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category	
19	Himalayan Restaurant	Filipino Restaurant	American Restaurant	
79	Himalayan Restaurant	Filipino Restaurant	American Restaurant	
113	Himalayan Restaurant	Filipino Restaurant	American Restaurant	
138	Himalayan Restaurant	Filipino Restaurant	American Restaurant	
247	Himalayan Restaurant	Filipino Restaurant	American Restaurant	

Cluster 8:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
12	City Island	American Restaurant	Hotpot Restaurant	
31	Westchester Square	American Restaurant	Hotpot Restaurant	
49	Greenpoint	American Restaurant	Hotpot Restaurant	
55	Crown Heights	American Restaurant	Hotpot Restaurant	
63	Bedford Stuyvesant	American Restaurant	Hotpot Restaurant	
67	Red Hook	American Restaurant	Hotpot Restaurant	
97	South Side	American Restaurant	Hotpot Restaurant	
105	Central Harlem	American Restaurant	Hotpot Restaurant	
111	Upper West Side	American Restaurant	Hotpot Restaurant	
115	Murray Hill	American Restaurant	Hotpot Restaurant	
117	Greenwich Village	American Restaurant	Hotpot Restaurant	
120	Tribeca	American Restaurant	Hotpot Restaurant	
127	Battery Park City	American Restaurant	Hotpot Restaurant	
128	Financial District	American Restaurant	Hotpot Restaurant	
139	Long Island City	American Restaurant	Hotpot Restaurant	
160	Jamaica Center	American Restaurant	Hotpot Restaurant	
180	Murray Hill	American Restaurant	Hotpot Restaurant	
233	Grant City	American Restaurant	Hotpot Restaurant	
271	Sutton Place	American Restaurant	Hotpot Restaurant	
276	Flatiron	American Restaurant	Hotpot Restaurant	
277	Sunnyside Gardens	American Restaurant	Hotpot Restaurant	
280	Vinegar Hill	American Restaurant	Hotpot Restaurant	
299	Kingsbridge Heights	American Restaurant	Hotpot Restaurant	
	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category	
12	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
31	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
49	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
55	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
63	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
67	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
97	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
105	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
111	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
115	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
117	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
120	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
127	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
128	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
139	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
160	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
180	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
233	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
271	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
276	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	
277	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant	

280	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant
299	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant

Cluster 9:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
114	Midtown	American Restaurant	Hotpot Restaurant	

	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category
114	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant

Cluster 10:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
289	Homecrest	Asian Restaurant	Hotpot Restaurant	

	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category
289	Himalayan Restaurant	Filipino Restaurant	American Restaurant

Cluster 11:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
61	Williamsburg	American Restaurant	Hotpot Restaurant	
250	Midtown South	American Restaurant	Hotpot Restaurant	
279	Fulton Ferry	American Restaurant	Hotpot Restaurant	

	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category
61	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant
250	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant
279	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant

Cluster 12:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
86	Downtown	Asian Restaurant	American Restaurant	
126	Gramercy	Asian Restaurant	American Restaurant	

	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category
86	Hotpot Restaurant	Himalayan Restaurant	Filipino Restaurant
126	Hotpot Restaurant	Himalayan Restaurant	Filipino Restaurant

Cluster 13:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
130	Woodside	Himalayan Restaurant	Hotpot Restaurant	

	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category
130	Filipino Restaurant	Asian Restaurant	American Restaurant

Cluster 14:

	Neighborhood	1st Top Venue Category	2nd Top Venue Category	\
102	Inwood	American Restaurant	Hotpot Restaurant	

	3rd Top Venue Category	4th Top Venue Category	5th Top Venue Category
102	Himalayan Restaurant	Filipino Restaurant	Asian Restaurant