

final_assignment

January 9, 2020

```
[2]: import numpy as np # library to handle data in a vectorized manner

import pandas as pd # library for data analysis
pd.set_option('display.max_columns', None)
pd.set_option('display.max_rows', None)

import json # library to handle JSON files

#!conda install -c conda-forge geopy --yes # uncomment this line if you haven't
    ↳ completed the Foursquare API lab
from geopy.geocoders import Nominatim # convert an address into latitude and
    ↳ longitude values

import requests # library to handle requests
from pandas.io.json import json_normalize # transform JSON file into a pandas
    ↳ dataframe

# Matplotlib and associated plotting modules
import matplotlib.cm as cm
import matplotlib.colors as colors

# import k-means from clustering stage
from sklearn.cluster import KMeans

#!conda install -c conda-forge folium=0.5.0 --yes # uncomment this line if you
    ↳ haven't completed the Foursquare API lab
import folium # map rendering library

print('Libraries imported.')
```

Libraries imported.

```
[2]: !wget -q -O 'newyork_data.json' https://geo.nyu.edu/download/file/
    ↳ nyu-2451-34572-geojson.json
print('Data downloaded!')
```

Data downloaded!

```
[3]: with open('newyork_data.json') as json_data:
      newyork_data = json.load(json_data)

      neighborhoods_data = newyork_data['features']

[4]: # define the dataframe columns
      column_names = ['Borough', 'Neighborhood', 'Latitude', 'Longitude']

      # instantiate the dataframe
      neighborhoods = pd.DataFrame(columns=column_names)

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

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

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

[6]: #!conda install -c conda-forge geopy --yes
      from geopy.geocoders import Nominatim

[7]: address = 'New York City, NY'

      geolocator = Nominatim(user_agent="ny_explorer")
      location = geolocator.geocode(address)
      latitude = location.latitude
      longitude = location.longitude
      print('The geographical coordinate of New York City are {}, {}'.format(
          latitude, longitude))
```

The geographical coordinate of New York City are 40.7127281, -74.0060152.

```
[8]: # create map of New York using latitude and longitude values
      map_newyork = folium.Map(location=[latitude, longitude], zoom_start=10)

      # add markers to map
      for lat, lng, borough, neighborhood in zip(neighborhoods['Latitude'],
          neighborhoods['Longitude'], neighborhoods['Borough'],
          neighborhoods['Neighborhood']):
          label = '{} {}'.format(neighborhood, borough)
```

```

label = folium.Popup(label, parse_html=True)
color = 'blue'
if borough=='Manhattan':
    color='red'
folium.CircleMarker(
    [lat, lng],
    radius=3,

    popup=label,
    color=None,
    fill=True,
    fill_color=color,
    fill_opacity=0.8,
    parse_html=False).add_to(map_newyork)
map_newyork.save('map.html')
map_newyork

```

[8]: <folium.folium.Map at 0x7f6d56ad2be0>

```

[9]: manhattan_data = neighborhoods[neighborhoods['Borough'] == 'Manhattan'].
    ↪reset_index(drop=True)
manhattan_data.head()

```

```

[9]:
   Borough      Neighborhood  Latitude  Longitude
0  Manhattan      Marble Hill  40.876551 -73.910660
1  Manhattan      Chinatown   40.715618 -73.994279
2  Manhattan  Washington Heights  40.851903 -73.936900
3  Manhattan      Inwood     40.867684 -73.921210
4  Manhattan  Hamilton Heights  40.823604 -73.949688

```

```

[11]: address = 'Manhattan, NY'

geolocator = Nominatim(user_agent="ny_explorer")
location = geolocator.geocode(address)
latitude = location.latitude
longitude = location.longitude
print('The geographical coordinate of Manhattan are {}, {}'.format(latitude,
    ↪longitude))

```

The geographical coordinate of Manhattan are 40.7896239, -73.9598939.

```

[12]: import os
CLIENT_ID = os.environ['foursquare_ClientId'] # your Foursquare ID
CLIENT_SECRET = os.environ['foursquare_ClientSecret'] # your Foursquare Secret
VERSION = '20180604'
LIMIT = 50

```

```

[13]: neighborhood_latitude = manhattan_data.loc[0, 'Latitude'] # neighborhood
    ↪latitude value

```

```

neighborhood_longitude = manhattan_data.loc[0, 'Longitude'] # neighborhood
    ↳ longitude value

neighborhood_name = manhattan_data.loc[0, 'Neighborhood'] # neighborhood name

print('Latitude and longitude values of {} are {}, {}.'.
    ↳ format(neighborhood_name,
                                                    ↳
    ↳ neighborhood_latitude,
                                                    ↳
    ↳ neighborhood_longitude))

```

Latitude and longitude values of Marble Hill are 40.87655077879964,
-73.91065965862981.

```

[14]: # type your answer here
radius = 500 # define radius
url = 'https://api.foursquare.com/v2/venues/explore?
    ↳ &client_id={} &client_secret={} &v={} &ll={}, {} &radius={} &limit={} &section=trending &categoryId=
    ↳ format(
        CLIENT_ID,
        CLIENT_SECRET,
        VERSION,
        neighborhood_latitude,
        neighborhood_longitude,
        radius,
        LIMIT)

```

```

[15]: results = requests.get(url).json()

```

```

[16]: # function that extracts the category of the venue
def get_category_type(row):
    try:
        categories_list = row['categories']
    except:
        categories_list = row['venue.categories']

    if len(categories_list) == 0:
        return None
    else:
        return categories_list[0]['shortName']

```

```

[17]: venues = results['response']['groups'][0]['items']

```

```

[18]: venues = results['response']['groups'][0]['items']

nearby_venues = json_normalize(venues) # flatten JSON

# filter columns

```

```

filtered_columns = ['venue.name', 'venue.categories', 'venue.location.lat',
    → 'venue.location.lng']
nearby_venues = nearby_venues.loc[:, filtered_columns]

# filter the category for each row
nearby_venues['venue.categories'] = nearby_venues.apply(get_category_type,
    → axis=1)

# clean columns
nearby_venues.columns = [col.split(".")[1] for col in nearby_venues.columns]
nearby_venues = np.round(nearby_venues, 2)
nearby_venues.head(5)

```

```

[18]:
      name  categories  lat  lng
0  Arturo's    Pizza  40.87 -73.91
1  Pick Up Six: Asian Kitchen    Asian  40.88 -73.91
2    Boston Market    American  40.88 -73.91
3        SUBWAY    Sandwiches  40.88 -73.91
4      Dunkin'    Donuts  40.88 -73.91

```

```

[19]: with open('marble_hill_venues.tex', 'w') as f:
      f.write(nearby_venues.head().to_latex(index=False))

```

```

[20]: print('{} venues were returned by Foursquare.'.format(nearby_venues.shape[0]))

```

13 venues were returned by Foursquare.

```

[33]: def getNearbyVenues(names, latitudes, longitudes, radius=500):

      venues_list=[]
      for name, lat, lng in zip(names, latitudes, longitudes):
          print(name)

          # create the API request URL
          url = 'https://api.foursquare.com/v2/venues/explore?
    → &client_id={} &client_secret={} &v={} &ll={}, {} &radius={} &limit={} &section=trending &categoryId=
    → format(
              CLIENT_ID,
              CLIENT_SECRET,
              VERSION,
              lat,
              lng,
              radius,
              LIMIT)

          # make the GET request
          results = requests.get(url).json()["response"]["groups"][0]["items"]

```

```

        # return only relevant information for each nearby venue
        venues_list.append([
            name,
            lat,
            lng,
            v['venue']['name'],
            v['venue']['location']['lat'],
            v['venue']['location']['lng'],
            v['venue']['categories'][0]['shortName']) for v in results])

    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',
                             'Venue Latitude',
                             'Venue Longitude',
                             'Venue Category']

    return(nearby_venues)

```

[34]: # type your answer here

```

manhattan_venues = getNearbyVenues(names=manhattan_data['Neighborhood'],
                                   latitudes=manhattan_data['Latitude'],
                                   longitudes=manhattan_data['Longitude']
                                   )

```

Marble Hill
 Chinatown
 Washington Heights
 Inwood
 Hamilton Heights
 Manhattanville
 Central Harlem
 East Harlem
 Upper East Side
 Yorkville
 Lenox Hill
 Roosevelt Island
 Upper West Side
 Lincoln Square
 Clinton
 Midtown
 Murray Hill
 Chelsea
 Greenwich Village

East Village
 Lower East Side
 Tribeca
 Little Italy
 Soho
 West Village
 Manhattan Valley
 Morningside Heights
 Gramercy
 Battery Park City
 Financial District
 Carnegie Hill
 Noho
 Civic Center
 Midtown South
 Sutton Place
 Turtle Bay
 Tudor City
 Stuyvesant Town
 Flatiron
 Hudson Yards

```
[35]: print(manhattan_venues.shape)
      manhattan_venues.head()
```

(1830, 7)

```
[35]: Neighborhood Neighborhood Latitude Neighborhood Longitude \
0 Marble Hill 40.876551 -73.91066
1 Marble Hill 40.876551 -73.91066
2 Marble Hill 40.876551 -73.91066
3 Marble Hill 40.876551 -73.91066
4 Marble Hill 40.876551 -73.91066
```

	Venue	Venue Latitude	Venue Longitude	Venue Category
0	Arturo's	40.874412	-73.910271	Pizza
1	Pick Up Six: Asian Kitchen	40.878075	-73.907033	Asian
2	Boston Market	40.877430	-73.905412	American
3	SUBWAY	40.878493	-73.905385	Sandwiches
4	Dunkin'	40.877136	-73.906666	Donuts

```
[36]: manhattan_venues.groupby('Neighborhood').count()
```

```
[36]: Neighborhood Latitude Neighborhood Longitude Venue \
Neighborhood
Battery Park City 32 32 32
Carnegie Hill 50 50 50
Central Harlem 46 46 46
```

Chelsea	50	50	50
Chinatown	50	50	50
Civic Center	50	50	50
Clinton	50	50	50
East Harlem	49	49	49
East Village	50	50	50
Financial District	50	50	50
Flatiron	50	50	50
Gramercy	50	50	50
Greenwich Village	50	50	50
Hamilton Heights	50	50	50
Hudson Yards	45	45	45
Inwood	49	49	49
Lenox Hill	50	50	50
Lincoln Square	50	50	50
Little Italy	50	50	50
Lower East Side	48	48	48
Manhattan Valley	44	44	44
Manhattanville	45	45	45
Marble Hill	13	13	13
Midtown	50	50	50
Midtown South	50	50	50
Morningside Heights	42	42	42
Murray Hill	50	50	50
Noho	50	50	50
Roosevelt Island	12	12	12
Soho	50	50	50
Stuyvesant Town	5	5	5
Sutton Place	50	50	50
Tribeca	50	50	50
Tudor City	50	50	50
Turtle Bay	50	50	50
Upper East Side	50	50	50
Upper West Side	50	50	50
Washington Heights	50	50	50
West Village	50	50	50
Yorkville	50	50	50

	Venue Latitude	Venue Longitude	Venue Category
Neighborhood			
Battery Park City	32	32	32
Carnegie Hill	50	50	50
Central Harlem	46	46	46
Chelsea	50	50	50
Chinatown	50	50	50
Civic Center	50	50	50
Clinton	50	50	50

East Harlem	49	49	49
East Village	50	50	50
Financial District	50	50	50
Flatiron	50	50	50
Gramercy	50	50	50
Greenwich Village	50	50	50
Hamilton Heights	50	50	50
Hudson Yards	45	45	45
Inwood	49	49	49
Lenox Hill	50	50	50
Lincoln Square	50	50	50
Little Italy	50	50	50
Lower East Side	48	48	48
Manhattan Valley	44	44	44
Manhattanville	45	45	45
Marble Hill	13	13	13
Midtown	50	50	50
Midtown South	50	50	50
Morningside Heights	42	42	42
Murray Hill	50	50	50
Noho	50	50	50
Roosevelt Island	12	12	12
Soho	50	50	50
Stuyvesant Town	5	5	5
Sutton Place	50	50	50
Tribeca	50	50	50
Tudor City	50	50	50
Turtle Bay	50	50	50
Upper East Side	50	50	50
Upper West Side	50	50	50
Washington Heights	50	50	50
West Village	50	50	50
Yorkville	50	50	50

```
[25]: print('There are {} uniques categories.'.format(len(manhattan_venues['Venue_
→Category'].unique())))
```

There are 104 uniques categories.

```
[37]: # one hot encoding
manhattan_onehot = pd.get_dummies(manhattan_venues[['Venue Category']],
→prefix="", prefix_sep="")

# add neighborhood column back to dataframe
manhattan_onehot['Neighborhood'] = manhattan_venues['Neighborhood']

# move neighborhood column to the first column
```

```
fixed_columns = [manhattan_onehot.columns[-1]] + list(manhattan_onehot.columns[:
→-1])
manhattan_onehot = manhattan_onehot[fixed_columns]

manhattan_onehot.head()
```

```
[37]: Neighborhood African American Argentinian Asian Australian Austrian \
0 Marble Hill 0 0 0 0 0 0
1 Marble Hill 0 0 0 1 0 0
2 Marble Hill 0 1 0 0 0 0
3 Marble Hill 0 0 0 0 0 0
4 Marble Hill 0 0 0 0 0 0
```

```
BBQ Bagels Bakery Belgian Bistro Brazilian Breakfast Burgers \
0 0 0 0 0 0 0 0 0
1 0 0 0 0 0 0 0 0
2 0 0 0 0 0 0 0 0
3 0 0 0 0 0 0 0 0
4 0 0 0 0 0 0 0 0
```

```
Burritos Cafeteria Café Cajun / Creole Caribbean Caucasian Chinese \
0 0 0 0 0 0 0 0
1 0 0 0 0 0 0 0
2 0 0 0 0 0 0 0
3 0 0 0 0 0 0 0
4 0 0 0 0 0 0 0
```

```
Creperie Cuban Czech Deli / Bodega Dim Sum Diner Donuts Dumplings \
0 0 0 0 0 0 0 0 0
1 0 0 0 0 0 0 0 0
2 0 0 0 0 0 0 0 0
3 0 0 0 0 0 0 0 0
4 0 0 0 0 0 0 1 0
```

```
Eastern European Egyptian Restaurant Empanada English Ethiopian \
0 0 0 0 0 0 0
1 0 0 0 0 0 0
2 0 0 0 0 0 0
3 0 0 0 0 0 0
4 0 0 0 0 0 0
```

```
Falafel Fast Food Filipino Food Food Court Food Truck French \
0 0 0 0 0 0 0 0
1 0 0 0 0 0 0 0
2 0 0 0 0 0 0 0
3 0 0 0 0 0 0 0
4 0 0 0 0 0 0 0
```

	Fried Chicken	Gastropub	German	Gluten-free	Greek	Hawaiian	Himalayan	\
0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	
2	0	0	0	0	0	0	0	
3	0	0	0	0	0	0	0	
4	0	0	0	0	0	0	0	

	Hot Dogs	Hotpot	Indian	Indonesian	Irish	Israeli	Italian	Japanese	\
0	0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	0	
2	0	0	0	0	0	0	0	0	
3	0	0	0	0	0	0	0	0	
4	0	0	0	0	0	0	0	0	

	Japanese Curry	Korean	Kosher	Latin American	Mac & Cheese	Malay	\
0	0	0	0	0	0	0	
1	0	0	0	0	0	0	
2	0	0	0	0	0	0	
3	0	0	0	0	0	0	
4	0	0	0	0	0	0	

	Mediterranean	Mexican	Middle Eastern	Modern European	\
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	0	0	0	

	Molecular Gastronomy	Moroccan	New American	Noodles	North Indian	\
0	0	0	0	0	0	
1	0	0	0	0	0	
2	0	0	0	0	0	
3	0	0	0	0	0	
4	0	0	0	0	0	

	Paella	Pakistani	Peruvian	Pet Café	Pizza	Poke Place	Ramen	\
0	0	0	0	0	1	0	0	
1	0	0	0	0	0	0	0	
2	0	0	0	0	0	0	0	
3	0	0	0	0	0	0	0	
4	0	0	0	0	0	0	0	

	Restaurant	Russian	Salad	Sandwiches	Scandinavian	Seafood	Shanghai	\
0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	
2	0	0	0	0	0	0	0	

3	0	0	0	1	0	0	0
4	0	0	0	0	0	0	0

	Snacks	Soup	South Indian	Southern / Soul	Spanish	Steakhouse	Sushi \
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
2	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0

	Szechuan	Tacos	Taiwanese	Tapas	Tex-Mex	Thai	Tonkatsu	Turkish	Udon \
0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0
2	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0

	Vegetarian / Vegan	Vietnamese	Wings
0	0	0	0
1	0	0	0
2	0	0	0
3	0	0	0
4	0	0	0

[38]: manhattan_onehot.shape

[38]: (1830, 105)

[39]: manhattan_grouped = manhattan_onehot.groupby('Neighborhood').mean().
 ↪reset_index()
 manhattan_grouped

[39]:

	Neighborhood	African	American	Argentinian	Asian \
0	Battery Park City	0.000000	0.031250	0.000000	0.000000
1	Carnegie Hill	0.000000	0.000000	0.020000	0.000000
2	Central Harlem	0.065217	0.043478	0.000000	0.000000
3	Chelsea	0.000000	0.060000	0.000000	0.020000
4	Chinatown	0.000000	0.000000	0.000000	0.040000
5	Civic Center	0.000000	0.080000	0.000000	0.000000
6	Clinton	0.000000	0.040000	0.000000	0.020000
7	East Harlem	0.000000	0.000000	0.000000	0.000000
8	East Village	0.000000	0.060000	0.020000	0.000000
9	Financial District	0.000000	0.080000	0.000000	0.000000
10	Flatiron	0.000000	0.020000	0.000000	0.020000
11	Gramercy	0.000000	0.060000	0.000000	0.000000
12	Greenwich Village	0.000000	0.020000	0.000000	0.000000
13	Hamilton Heights	0.000000	0.000000	0.000000	0.000000
14	Hudson Yards	0.000000	0.111111	0.000000	0.022222
15	Inwood	0.000000	0.040816	0.000000	0.000000

16	Lenox Hill	0.000000	0.000000	0.000000	0.000000
17	Lincoln Square	0.000000	0.100000	0.000000	0.000000
18	Little Italy	0.000000	0.020000	0.000000	0.020000
19	Lower East Side	0.000000	0.020833	0.020833	0.000000
20	Manhattan Valley	0.000000	0.022727	0.000000	0.000000
21	Manhattanville	0.000000	0.022222	0.000000	0.000000
22	Marble Hill	0.000000	0.076923	0.000000	0.076923
23	Midtown	0.000000	0.060000	0.000000	0.000000
24	Midtown South	0.000000	0.000000	0.000000	0.000000
25	Morningside Heights	0.000000	0.071429	0.000000	0.000000
26	Murray Hill	0.000000	0.120000	0.000000	0.020000
27	Noho	0.000000	0.040000	0.020000	0.040000
28	Roosevelt Island	0.000000	0.000000	0.000000	0.000000
29	Soho	0.000000	0.020000	0.000000	0.020000
30	Stuyvesant Town	0.000000	0.000000	0.000000	0.000000
31	Sutton Place	0.000000	0.100000	0.000000	0.020000
32	Tribeca	0.000000	0.100000	0.020000	0.020000
33	Tudor City	0.000000	0.060000	0.000000	0.020000
34	Turtle Bay	0.000000	0.000000	0.000000	0.020000
35	Upper East Side	0.000000	0.080000	0.000000	0.020000
36	Upper West Side	0.000000	0.040000	0.000000	0.020000
37	Washington Heights	0.000000	0.020000	0.000000	0.000000
38	West Village	0.000000	0.060000	0.000000	0.000000
39	Yorkville	0.000000	0.000000	0.000000	0.000000

	Australian	Austrian	BBQ	Bagels	Bakery	Belgian	Bistro	\
0	0.00	0.00	0.062500	0.000000	0.031250	0.00	0.031250	
1	0.00	0.00	0.000000	0.000000	0.120000	0.00	0.000000	
2	0.00	0.00	0.021739	0.021739	0.021739	0.00	0.000000	
3	0.00	0.00	0.000000	0.020000	0.100000	0.00	0.000000	
4	0.00	0.00	0.000000	0.000000	0.060000	0.00	0.000000	
5	0.00	0.00	0.000000	0.000000	0.060000	0.00	0.000000	
6	0.00	0.00	0.000000	0.000000	0.000000	0.00	0.000000	
7	0.00	0.00	0.000000	0.000000	0.102041	0.00	0.000000	
8	0.00	0.00	0.000000	0.000000	0.000000	0.00	0.000000	
9	0.00	0.00	0.000000	0.020000	0.020000	0.00	0.000000	
10	0.00	0.00	0.000000	0.000000	0.040000	0.02	0.000000	
11	0.00	0.00	0.000000	0.060000	0.020000	0.00	0.000000	
12	0.02	0.00	0.000000	0.020000	0.040000	0.00	0.000000	
13	0.00	0.00	0.000000	0.000000	0.020000	0.00	0.000000	
14	0.00	0.00	0.022222	0.000000	0.000000	0.00	0.000000	
15	0.00	0.00	0.000000	0.020408	0.061224	0.00	0.020408	
16	0.00	0.00	0.020000	0.020000	0.020000	0.00	0.000000	
17	0.00	0.00	0.000000	0.020000	0.020000	0.00	0.000000	
18	0.02	0.00	0.000000	0.000000	0.100000	0.00	0.000000	
19	0.00	0.00	0.000000	0.020833	0.062500	0.00	0.000000	
20	0.00	0.00	0.000000	0.000000	0.022727	0.00	0.000000	

21	0.00	0.00	0.022222	0.000000	0.044444	0.00	0.000000
22	0.00	0.00	0.000000	0.000000	0.076923	0.00	0.000000
23	0.00	0.00	0.000000	0.000000	0.020000	0.00	0.000000
24	0.00	0.00	0.000000	0.020000	0.100000	0.00	0.000000
25	0.00	0.00	0.000000	0.000000	0.000000	0.00	0.000000
26	0.00	0.00	0.000000	0.040000	0.040000	0.00	0.000000
27	0.02	0.00	0.000000	0.020000	0.020000	0.00	0.000000
28	0.00	0.00	0.000000	0.000000	0.000000	0.00	0.000000
29	0.02	0.00	0.000000	0.020000	0.080000	0.00	0.000000
30	0.00	0.00	0.000000	0.000000	0.000000	0.00	0.200000
31	0.00	0.00	0.000000	0.100000	0.020000	0.00	0.000000
32	0.00	0.00	0.000000	0.000000	0.060000	0.00	0.000000
33	0.00	0.00	0.000000	0.020000	0.000000	0.00	0.000000
34	0.00	0.00	0.000000	0.000000	0.000000	0.00	0.000000
35	0.00	0.00	0.000000	0.040000	0.020000	0.00	0.000000
36	0.02	0.00	0.000000	0.020000	0.060000	0.00	0.000000
37	0.00	0.00	0.000000	0.000000	0.040000	0.00	0.000000
38	0.00	0.02	0.020000	0.000000	0.020000	0.00	0.020000
39	0.00	0.00	0.000000	0.020000	0.040000	0.00	0.020000

	Brazilian	Breakfast	Burgers	Burritos	Cafeteria	Café \
0	0.00	0.000000	0.062500	0.03125	0.000000	0.031250
1	0.00	0.020000	0.000000	0.00000	0.000000	0.100000
2	0.00	0.021739	0.021739	0.00000	0.021739	0.021739
3	0.00	0.020000	0.000000	0.00000	0.000000	0.080000
4	0.00	0.020000	0.000000	0.00000	0.000000	0.000000
5	0.00	0.000000	0.000000	0.02000	0.000000	0.060000
6	0.02	0.020000	0.020000	0.00000	0.000000	0.060000
7	0.00	0.000000	0.040816	0.00000	0.000000	0.061224
8	0.00	0.000000	0.000000	0.00000	0.000000	0.020000
9	0.00	0.000000	0.000000	0.00000	0.000000	0.080000
10	0.00	0.000000	0.020000	0.04000	0.000000	0.020000
11	0.00	0.000000	0.000000	0.00000	0.000000	0.040000
12	0.00	0.000000	0.000000	0.00000	0.000000	0.080000
13	0.00	0.000000	0.020000	0.00000	0.000000	0.080000
14	0.00	0.022222	0.044444	0.00000	0.000000	0.088889
15	0.00	0.000000	0.020408	0.00000	0.000000	0.081633
16	0.00	0.020000	0.040000	0.00000	0.000000	0.040000
17	0.00	0.000000	0.020000	0.00000	0.000000	0.160000
18	0.00	0.020000	0.000000	0.00000	0.000000	0.080000
19	0.00	0.000000	0.000000	0.00000	0.000000	0.041667
20	0.00	0.000000	0.022727	0.00000	0.000000	0.045455
21	0.00	0.000000	0.022222	0.00000	0.000000	0.022222
22	0.00	0.000000	0.000000	0.00000	0.000000	0.000000
23	0.00	0.000000	0.040000	0.02000	0.000000	0.020000
24	0.00	0.000000	0.020000	0.00000	0.000000	0.020000
25	0.00	0.000000	0.071429	0.00000	0.000000	0.071429

26	0.00	0.000000	0.020000	0.000000	0.000000	0.040000
27	0.00	0.020000	0.020000	0.000000	0.000000	0.020000
28	0.00	0.000000	0.000000	0.000000	0.000000	0.083333
29	0.00	0.000000	0.000000	0.000000	0.000000	0.120000
30	0.00	0.000000	0.000000	0.000000	0.000000	0.000000
31	0.00	0.000000	0.040000	0.020000	0.000000	0.020000
32	0.00	0.000000	0.060000	0.000000	0.000000	0.060000
33	0.00	0.000000	0.020000	0.000000	0.000000	0.120000
34	0.00	0.000000	0.000000	0.000000	0.000000	0.120000
35	0.00	0.000000	0.060000	0.000000	0.000000	0.000000
36	0.00	0.040000	0.000000	0.020000	0.000000	0.040000
37	0.00	0.020000	0.000000	0.000000	0.000000	0.040000
38	0.02	0.040000	0.040000	0.000000	0.000000	0.040000
39	0.00	0.000000	0.020000	0.000000	0.000000	0.020000

	Cajun / Creole	Caribbean	Caucasian	Chinese	Creperie	Cuban	Czech \
0	0.00	0.000000	0.000000	0.093750	0.00	0.000000	0.00
1	0.00	0.000000	0.000000	0.040000	0.00	0.000000	0.00
2	0.00	0.065217	0.000000	0.086957	0.00	0.000000	0.00
3	0.00	0.020000	0.000000	0.020000	0.02	0.000000	0.00
4	0.00	0.000000	0.000000	0.240000	0.00	0.000000	0.00
5	0.02	0.000000	0.000000	0.000000	0.00	0.020000	0.00
6	0.00	0.020000	0.020000	0.000000	0.02	0.000000	0.00
7	0.00	0.000000	0.000000	0.020408	0.00	0.020408	0.00
8	0.00	0.000000	0.000000	0.020000	0.00	0.000000	0.00
9	0.00	0.000000	0.000000	0.020000	0.00	0.020000	0.00
10	0.00	0.000000	0.000000	0.000000	0.00	0.000000	0.00
11	0.00	0.000000	0.000000	0.000000	0.00	0.000000	0.00
12	0.00	0.000000	0.000000	0.000000	0.00	0.000000	0.00
13	0.00	0.040000	0.000000	0.080000	0.00	0.000000	0.00
14	0.00	0.000000	0.022222	0.000000	0.00	0.000000	0.00
15	0.00	0.040816	0.000000	0.040816	0.00	0.000000	0.00
16	0.00	0.000000	0.000000	0.040000	0.00	0.020000	0.02
17	0.00	0.000000	0.000000	0.040000	0.00	0.000000	0.00
18	0.00	0.000000	0.000000	0.080000	0.00	0.000000	0.00
19	0.00	0.000000	0.000000	0.083333	0.00	0.000000	0.00
20	0.00	0.022727	0.000000	0.045455	0.00	0.000000	0.00
21	0.00	0.022222	0.000000	0.111111	0.00	0.022222	0.00
22	0.00	0.000000	0.000000	0.000000	0.00	0.000000	0.00
23	0.00	0.000000	0.000000	0.020000	0.02	0.000000	0.00
24	0.00	0.020000	0.000000	0.000000	0.00	0.000000	0.00
25	0.00	0.000000	0.000000	0.047619	0.00	0.000000	0.00
26	0.00	0.000000	0.000000	0.040000	0.00	0.000000	0.00
27	0.00	0.000000	0.000000	0.000000	0.00	0.000000	0.00
28	0.00	0.000000	0.000000	0.083333	0.00	0.000000	0.00
29	0.00	0.000000	0.000000	0.020000	0.00	0.000000	0.00
30	0.00	0.000000	0.000000	0.000000	0.00	0.000000	0.00

31	0.00	0.000000	0.000000	0.060000	0.00	0.000000	0.00
32	0.00	0.000000	0.000000	0.040000	0.00	0.000000	0.00
33	0.00	0.000000	0.000000	0.040000	0.00	0.000000	0.00
34	0.00	0.000000	0.000000	0.000000	0.00	0.000000	0.00
35	0.00	0.000000	0.000000	0.020000	0.00	0.000000	0.00
36	0.00	0.000000	0.000000	0.020000	0.00	0.000000	0.00
37	0.00	0.020000	0.000000	0.060000	0.00	0.000000	0.00
38	0.00	0.000000	0.000000	0.000000	0.00	0.000000	0.00
39	0.00	0.000000	0.000000	0.060000	0.00	0.000000	0.00

	Deli / Bodega	Dim Sum	Diner	Donuts	Dumplings	Eastern European	\
0	0.000000	0.000000	0.000000	0.031250	0.000000		0.00
1	0.020000	0.000000	0.040000	0.000000	0.000000		0.00
2	0.108696	0.000000	0.000000	0.000000	0.000000		0.00
3	0.020000	0.000000	0.000000	0.000000	0.000000		0.00
4	0.040000	0.020000	0.000000	0.000000	0.080000		0.00
5	0.040000	0.000000	0.040000	0.000000	0.000000		0.00
6	0.160000	0.000000	0.000000	0.020000	0.000000		0.00
7	0.102041	0.000000	0.000000	0.020408	0.000000		0.00
8	0.040000	0.000000	0.000000	0.000000	0.020000		0.02
9	0.040000	0.000000	0.000000	0.020000	0.000000		0.00
10	0.000000	0.000000	0.000000	0.020000	0.000000		0.00
11	0.100000	0.000000	0.060000	0.000000	0.000000		0.00
12	0.000000	0.000000	0.000000	0.000000	0.000000		0.00
13	0.140000	0.000000	0.000000	0.060000	0.000000		0.00
14	0.044444	0.000000	0.000000	0.000000	0.000000		0.00
15	0.040816	0.000000	0.040816	0.040816	0.020408		0.00
16	0.100000	0.000000	0.020000	0.000000	0.000000		0.00
17	0.040000	0.020000	0.020000	0.000000	0.000000		0.00
18	0.000000	0.000000	0.000000	0.000000	0.020000		0.00
19	0.104167	0.020833	0.041667	0.000000	0.000000		0.00
20	0.045455	0.000000	0.022727	0.000000	0.000000		0.00
21	0.088889	0.000000	0.022222	0.022222	0.022222		0.00
22	0.153846	0.000000	0.076923	0.076923	0.000000		0.00
23	0.080000	0.000000	0.000000	0.000000	0.000000		0.00
24	0.080000	0.000000	0.000000	0.000000	0.000000		0.00
25	0.119048	0.000000	0.023810	0.000000	0.000000		0.00
26	0.020000	0.000000	0.020000	0.000000	0.000000		0.00
27	0.020000	0.000000	0.020000	0.000000	0.000000		0.00
28	0.166667	0.000000	0.000000	0.000000	0.000000		0.00
29	0.000000	0.000000	0.000000	0.000000	0.000000		0.00
30	0.200000	0.000000	0.000000	0.000000	0.000000		0.00
31	0.020000	0.000000	0.020000	0.000000	0.000000		0.02
32	0.060000	0.000000	0.020000	0.000000	0.000000		0.00
33	0.120000	0.000000	0.040000	0.000000	0.000000		0.00
34	0.080000	0.000000	0.020000	0.000000	0.000000		0.00
35	0.000000	0.000000	0.060000	0.020000	0.000000		0.00

36	0.000000	0.000000	0.000000	0.000000	0.000000	0.00
37	0.180000	0.000000	0.020000	0.020000	0.000000	0.00
38	0.040000	0.000000	0.000000	0.000000	0.000000	0.00
39	0.100000	0.000000	0.000000	0.000000	0.000000	0.00

	Egyptian Restaurant	Empanada	English	Ethiopian	Falafel	Fast Food \
0	0.00	0.000000	0.00	0.000000	0.000000	0.000000
1	0.00	0.000000	0.00	0.000000	0.000000	0.020000
2	0.00	0.000000	0.00	0.021739	0.000000	0.000000
3	0.00	0.000000	0.00	0.000000	0.000000	0.000000
4	0.00	0.000000	0.00	0.000000	0.000000	0.000000
5	0.00	0.000000	0.00	0.000000	0.020000	0.000000
6	0.00	0.000000	0.00	0.000000	0.000000	0.000000
7	0.00	0.000000	0.00	0.000000	0.000000	0.040816
8	0.00	0.000000	0.00	0.020000	0.000000	0.000000
9	0.00	0.000000	0.00	0.000000	0.020000	0.000000
10	0.00	0.000000	0.02	0.000000	0.000000	0.000000
11	0.00	0.000000	0.00	0.000000	0.000000	0.000000
12	0.00	0.000000	0.00	0.000000	0.000000	0.000000
13	0.00	0.000000	0.00	0.000000	0.020000	0.020000
14	0.00	0.000000	0.00	0.000000	0.000000	0.000000
15	0.00	0.020408	0.00	0.000000	0.000000	0.020408
16	0.00	0.000000	0.00	0.000000	0.000000	0.000000
17	0.00	0.000000	0.00	0.000000	0.000000	0.000000
18	0.02	0.000000	0.00	0.000000	0.000000	0.000000
19	0.00	0.000000	0.00	0.000000	0.000000	0.000000
20	0.00	0.000000	0.00	0.022727	0.022727	0.000000
21	0.00	0.000000	0.00	0.000000	0.022222	0.022222
22	0.00	0.000000	0.00	0.000000	0.000000	0.000000
23	0.00	0.000000	0.00	0.000000	0.000000	0.020000
24	0.00	0.000000	0.00	0.000000	0.000000	0.040000
25	0.00	0.000000	0.00	0.023810	0.023810	0.000000
26	0.00	0.000000	0.00	0.000000	0.000000	0.000000
27	0.00	0.000000	0.00	0.000000	0.000000	0.000000
28	0.00	0.000000	0.00	0.000000	0.000000	0.000000
29	0.02	0.000000	0.00	0.000000	0.000000	0.000000
30	0.00	0.000000	0.00	0.000000	0.000000	0.000000
31	0.00	0.000000	0.00	0.000000	0.000000	0.000000
32	0.00	0.000000	0.00	0.000000	0.000000	0.000000
33	0.00	0.000000	0.00	0.000000	0.000000	0.000000
34	0.00	0.000000	0.00	0.000000	0.000000	0.000000
35	0.00	0.000000	0.00	0.000000	0.000000	0.000000
36	0.00	0.000000	0.00	0.000000	0.000000	0.000000
37	0.00	0.000000	0.00	0.000000	0.000000	0.020000
38	0.00	0.000000	0.00	0.000000	0.000000	0.000000
39	0.00	0.000000	0.00	0.000000	0.000000	0.000000

	Filipino	Food	Food Court	Food Truck	French	Fried Chicken	\
0	0.000000	0.00	0.062500	0.000000	0.000000	0.000000	
1	0.000000	0.00	0.000000	0.000000	0.060000	0.000000	
2	0.000000	0.00	0.000000	0.000000	0.043478	0.086957	
3	0.000000	0.00	0.000000	0.000000	0.040000	0.000000	
4	0.000000	0.02	0.000000	0.000000	0.000000	0.000000	
5	0.000000	0.00	0.020000	0.000000	0.080000	0.000000	
6	0.000000	0.00	0.020000	0.020000	0.040000	0.020000	
7	0.000000	0.00	0.000000	0.000000	0.020408	0.020408	
8	0.040000	0.00	0.000000	0.000000	0.020000	0.000000	
9	0.000000	0.00	0.000000	0.060000	0.020000	0.020000	
10	0.000000	0.00	0.000000	0.000000	0.040000	0.000000	
11	0.020000	0.00	0.000000	0.020000	0.000000	0.000000	
12	0.000000	0.00	0.000000	0.040000	0.120000	0.000000	
13	0.000000	0.00	0.000000	0.020000	0.000000	0.040000	
14	0.000000	0.00	0.022222	0.022222	0.000000	0.000000	
15	0.000000	0.00	0.000000	0.000000	0.000000	0.000000	
16	0.000000	0.00	0.000000	0.020000	0.020000	0.000000	
17	0.000000	0.02	0.020000	0.140000	0.060000	0.000000	
18	0.000000	0.00	0.000000	0.000000	0.040000	0.000000	
19	0.020833	0.00	0.000000	0.000000	0.020833	0.020833	
20	0.000000	0.00	0.000000	0.000000	0.045455	0.022727	
21	0.000000	0.00	0.022222	0.022222	0.000000	0.044444	
22	0.000000	0.00	0.000000	0.000000	0.000000	0.000000	
23	0.000000	0.00	0.020000	0.080000	0.040000	0.000000	
24	0.000000	0.02	0.020000	0.000000	0.020000	0.020000	
25	0.000000	0.00	0.000000	0.142857	0.000000	0.000000	
26	0.000000	0.00	0.020000	0.000000	0.000000	0.020000	
27	0.000000	0.00	0.000000	0.000000	0.000000	0.000000	
28	0.000000	0.00	0.000000	0.000000	0.000000	0.000000	
29	0.000000	0.00	0.000000	0.000000	0.120000	0.000000	
30	0.000000	0.00	0.000000	0.000000	0.000000	0.000000	
31	0.000000	0.00	0.000000	0.040000	0.020000	0.020000	
32	0.000000	0.00	0.020000	0.020000	0.040000	0.000000	
33	0.000000	0.00	0.000000	0.060000	0.000000	0.000000	
34	0.000000	0.00	0.000000	0.080000	0.020000	0.000000	
35	0.000000	0.00	0.000000	0.020000	0.060000	0.000000	
36	0.000000	0.00	0.000000	0.000000	0.040000	0.000000	
37	0.000000	0.02	0.000000	0.040000	0.000000	0.000000	
38	0.000000	0.00	0.000000	0.020000	0.020000	0.000000	
39	0.000000	0.00	0.000000	0.040000	0.000000	0.000000	

	Gastropub	German	Gluten-free	Greek	Hawaiian	Himalayan	Hot Dogs	\
0	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00	
1	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.02	
2	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00	
3	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00	

4	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
5	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
6	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
7	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
8	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.02
9	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
10	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
11	0.000000	0.02	0.00	0.000000	0.000000	0.00	0.00
12	0.000000	0.00	0.02	0.000000	0.000000	0.00	0.00
13	0.020000	0.00	0.00	0.000000	0.000000	0.00	0.00
14	0.000000	0.00	0.00	0.044444	0.000000	0.00	0.00
15	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
16	0.000000	0.00	0.00	0.060000	0.000000	0.00	0.00
17	0.000000	0.00	0.00	0.020000	0.000000	0.00	0.00
18	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
19	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
20	0.022727	0.00	0.00	0.000000	0.022727	0.00	0.00
21	0.022222	0.00	0.00	0.000000	0.000000	0.00	0.00
22	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
23	0.020000	0.02	0.00	0.020000	0.000000	0.00	0.00
24	0.020000	0.02	0.00	0.000000	0.000000	0.00	0.00
25	0.000000	0.00	0.00	0.023810	0.000000	0.00	0.00
26	0.020000	0.00	0.00	0.020000	0.020000	0.00	0.00
27	0.000000	0.00	0.00	0.000000	0.000000	0.02	0.00
28	0.000000	0.00	0.00	0.083333	0.000000	0.00	0.00
29	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
30	0.000000	0.20	0.00	0.000000	0.000000	0.00	0.00
31	0.000000	0.00	0.00	0.020000	0.000000	0.00	0.00
32	0.000000	0.00	0.00	0.020000	0.000000	0.00	0.00
33	0.000000	0.00	0.00	0.060000	0.020000	0.00	0.00
34	0.000000	0.00	0.00	0.040000	0.000000	0.00	0.00
35	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
36	0.000000	0.00	0.00	0.020000	0.000000	0.00	0.00
37	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00
38	0.060000	0.00	0.00	0.020000	0.000000	0.00	0.00
39	0.000000	0.00	0.00	0.000000	0.000000	0.00	0.00

	Hotpot	Indian	Indonesian	Irish	Israeli	Italian	Japanese	\
0	0.00	0.000000	0.00	0.00	0.00	0.125000	0.000000	
1	0.00	0.040000	0.00	0.00	0.00	0.040000	0.040000	
2	0.00	0.000000	0.00	0.00	0.00	0.000000	0.000000	
3	0.00	0.020000	0.00	0.00	0.02	0.100000	0.060000	
4	0.02	0.000000	0.00	0.00	0.00	0.060000	0.000000	
5	0.00	0.040000	0.00	0.00	0.00	0.100000	0.000000	
6	0.00	0.020000	0.00	0.00	0.00	0.100000	0.000000	
7	0.00	0.000000	0.00	0.00	0.00	0.000000	0.000000	
8	0.00	0.000000	0.00	0.00	0.00	0.080000	0.100000	

9	0.00	0.020000	0.00	0.00	0.00	0.060000	0.060000
10	0.00	0.020000	0.00	0.00	0.02	0.160000	0.020000
11	0.00	0.000000	0.00	0.02	0.00	0.100000	0.000000
12	0.00	0.040000	0.00	0.00	0.00	0.180000	0.020000
13	0.00	0.040000	0.00	0.00	0.00	0.000000	0.020000
14	0.00	0.022222	0.00	0.00	0.00	0.088889	0.000000
15	0.00	0.000000	0.00	0.00	0.00	0.020408	0.000000
16	0.00	0.020000	0.00	0.00	0.00	0.160000	0.040000
17	0.00	0.020000	0.00	0.00	0.00	0.100000	0.000000
18	0.00	0.000000	0.02	0.00	0.00	0.140000	0.000000
19	0.00	0.000000	0.00	0.00	0.00	0.020833	0.041667
20	0.00	0.068182	0.00	0.00	0.00	0.022727	0.022727
21	0.00	0.022222	0.00	0.00	0.00	0.044444	0.000000
22	0.00	0.000000	0.00	0.00	0.00	0.000000	0.000000
23	0.00	0.040000	0.00	0.00	0.00	0.020000	0.040000
24	0.00	0.000000	0.00	0.00	0.00	0.020000	0.120000
25	0.00	0.023810	0.00	0.00	0.00	0.023810	0.000000
26	0.00	0.020000	0.00	0.00	0.00	0.040000	0.040000
27	0.00	0.020000	0.00	0.00	0.00	0.160000	0.080000
28	0.00	0.000000	0.00	0.00	0.00	0.000000	0.083333
29	0.00	0.020000	0.02	0.00	0.00	0.160000	0.000000
30	0.00	0.000000	0.00	0.00	0.00	0.000000	0.000000
31	0.00	0.040000	0.00	0.02	0.00	0.080000	0.040000
32	0.00	0.020000	0.00	0.00	0.00	0.100000	0.000000
33	0.00	0.000000	0.00	0.00	0.00	0.040000	0.000000
34	0.00	0.040000	0.00	0.00	0.00	0.120000	0.060000
35	0.00	0.000000	0.00	0.00	0.00	0.200000	0.020000
36	0.00	0.040000	0.00	0.00	0.02	0.140000	0.020000
37	0.00	0.000000	0.00	0.00	0.00	0.020000	0.000000
38	0.00	0.040000	0.00	0.00	0.00	0.140000	0.020000
39	0.00	0.040000	0.00	0.00	0.00	0.120000	0.040000

	Japanese Curry	Korean	Kosher	Latin American	Mac & Cheese	\
0	0.000000	0.000000	0.000000	0.000000	0.00	
1	0.000000	0.000000	0.020000	0.000000	0.00	
2	0.000000	0.000000	0.000000	0.000000	0.00	
3	0.000000	0.000000	0.000000	0.000000	0.00	
4	0.000000	0.020000	0.000000	0.000000	0.00	
5	0.000000	0.040000	0.000000	0.000000	0.00	
6	0.000000	0.020000	0.000000	0.000000	0.00	
7	0.000000	0.000000	0.000000	0.081633	0.00	
8	0.000000	0.000000	0.000000	0.000000	0.00	
9	0.000000	0.000000	0.000000	0.020000	0.00	
10	0.000000	0.060000	0.000000	0.000000	0.00	
11	0.000000	0.000000	0.000000	0.000000	0.00	
12	0.000000	0.020000	0.000000	0.000000	0.00	
13	0.000000	0.000000	0.000000	0.020000	0.00	

14	0.000000	0.000000	0.000000	0.000000	0.00
15	0.000000	0.000000	0.000000	0.040816	0.00
16	0.000000	0.000000	0.000000	0.000000	0.00
17	0.000000	0.000000	0.000000	0.000000	0.00
18	0.000000	0.020000	0.000000	0.000000	0.00
19	0.000000	0.000000	0.000000	0.041667	0.00
20	0.000000	0.022727	0.000000	0.022727	0.00
21	0.022222	0.000000	0.000000	0.000000	0.00
22	0.000000	0.000000	0.000000	0.000000	0.00
23	0.000000	0.000000	0.000000	0.000000	0.00
24	0.000000	0.160000	0.000000	0.000000	0.00
25	0.000000	0.047619	0.000000	0.000000	0.00
26	0.000000	0.000000	0.000000	0.020000	0.00
27	0.000000	0.000000	0.000000	0.020000	0.00
28	0.000000	0.000000	0.083333	0.000000	0.00
29	0.000000	0.020000	0.000000	0.020000	0.00
30	0.000000	0.000000	0.000000	0.000000	0.00
31	0.000000	0.000000	0.000000	0.020000	0.00
32	0.000000	0.020000	0.000000	0.000000	0.00
33	0.000000	0.000000	0.000000	0.020000	0.00
34	0.020000	0.000000	0.000000	0.000000	0.00
35	0.000000	0.000000	0.000000	0.000000	0.00
36	0.000000	0.000000	0.000000	0.020000	0.00
37	0.000000	0.000000	0.000000	0.060000	0.00
38	0.000000	0.000000	0.000000	0.020000	0.02
39	0.000000	0.000000	0.000000	0.020000	0.00

	Malay	Mediterranean	Mexican	Middle Eastern	Modern European	\
0	0.000000	0.031250	0.031250	0.00	0.00	
1	0.000000	0.020000	0.020000	0.00	0.00	
2	0.000000	0.000000	0.000000	0.00	0.00	
3	0.000000	0.020000	0.020000	0.00	0.00	
4	0.060000	0.000000	0.040000	0.00	0.00	
5	0.000000	0.000000	0.020000	0.00	0.02	
6	0.000000	0.000000	0.040000	0.02	0.00	
7	0.000000	0.000000	0.142857	0.00	0.00	
8	0.000000	0.000000	0.060000	0.04	0.00	
9	0.000000	0.020000	0.040000	0.02	0.00	
10	0.000000	0.020000	0.020000	0.00	0.00	
11	0.000000	0.000000	0.080000	0.00	0.00	
12	0.000000	0.000000	0.040000	0.00	0.00	
13	0.000000	0.020000	0.080000	0.00	0.00	
14	0.000000	0.000000	0.000000	0.00	0.00	
15	0.000000	0.000000	0.081633	0.00	0.00	
16	0.000000	0.000000	0.040000	0.02	0.00	
17	0.000000	0.040000	0.020000	0.00	0.00	
18	0.040000	0.040000	0.000000	0.02	0.00	

19	0.000000	0.020833	0.062500	0.00	0.00
20	0.022727	0.000000	0.113636	0.00	0.00
21	0.000000	0.000000	0.066667	0.00	0.00
22	0.000000	0.000000	0.000000	0.00	0.00
23	0.000000	0.040000	0.040000	0.00	0.00
24	0.000000	0.000000	0.000000	0.00	0.00
25	0.000000	0.023810	0.023810	0.00	0.00
26	0.000000	0.040000	0.040000	0.00	0.00
27	0.000000	0.020000	0.040000	0.00	0.00
28	0.000000	0.000000	0.000000	0.00	0.00
29	0.020000	0.060000	0.020000	0.00	0.00
30	0.000000	0.000000	0.000000	0.00	0.00
31	0.000000	0.000000	0.040000	0.00	0.00
32	0.000000	0.000000	0.000000	0.00	0.02
33	0.000000	0.020000	0.100000	0.00	0.00
34	0.000000	0.000000	0.000000	0.00	0.00
35	0.000000	0.020000	0.040000	0.00	0.00
36	0.000000	0.040000	0.040000	0.02	0.00
37	0.000000	0.000000	0.060000	0.00	0.00
38	0.000000	0.020000	0.000000	0.00	0.00
39	0.000000	0.000000	0.040000	0.00	0.00

	Molecular Gastronomy	Moroccan	New American	Noodles	North Indian \
0	0.00	0.00	0.000000	0.000000	0.00
1	0.00	0.00	0.040000	0.000000	0.00
2	0.00	0.00	0.000000	0.000000	0.00
3	0.00	0.00	0.000000	0.000000	0.00
4	0.00	0.00	0.000000	0.040000	0.00
5	0.02	0.00	0.020000	0.000000	0.00
6	0.00	0.00	0.020000	0.000000	0.00
7	0.00	0.00	0.000000	0.000000	0.00
8	0.00	0.02	0.000000	0.000000	0.00
9	0.00	0.00	0.020000	0.000000	0.00
10	0.00	0.00	0.040000	0.020000	0.00
11	0.00	0.00	0.000000	0.000000	0.00
12	0.00	0.00	0.020000	0.000000	0.00
13	0.00	0.00	0.000000	0.000000	0.00
14	0.00	0.00	0.022222	0.000000	0.00
15	0.00	0.00	0.000000	0.000000	0.00
16	0.00	0.00	0.000000	0.000000	0.00
17	0.00	0.00	0.000000	0.000000	0.00
18	0.00	0.00	0.000000	0.000000	0.00
19	0.00	0.00	0.020833	0.000000	0.00
20	0.00	0.00	0.000000	0.022727	0.00
21	0.00	0.00	0.000000	0.022222	0.00
22	0.00	0.00	0.000000	0.000000	0.00
23	0.00	0.00	0.000000	0.000000	0.00

24	0.00	0.00	0.020000	0.000000	0.00
25	0.00	0.00	0.047619	0.000000	0.00
26	0.00	0.00	0.020000	0.000000	0.00
27	0.00	0.00	0.000000	0.000000	0.00
28	0.00	0.00	0.000000	0.083333	0.00
29	0.00	0.00	0.000000	0.000000	0.00
30	0.00	0.00	0.000000	0.000000	0.00
31	0.00	0.00	0.000000	0.000000	0.00
32	0.00	0.00	0.020000	0.020000	0.00
33	0.00	0.00	0.000000	0.000000	0.00
34	0.00	0.00	0.000000	0.020000	0.00
35	0.00	0.00	0.000000	0.000000	0.02
36	0.00	0.00	0.020000	0.000000	0.00
37	0.00	0.00	0.020000	0.000000	0.00
38	0.00	0.00	0.020000	0.000000	0.00
39	0.00	0.00	0.040000	0.000000	0.00

	Paella	Pakistani	Peruvian	Pet Café	Pizza	Poke Place	Ramen	\
0	0.00	0.000000	0.000000	0.000000	0.125000	0.00	0.000000	
1	0.00	0.000000	0.000000	0.000000	0.120000	0.00	0.000000	
2	0.00	0.000000	0.000000	0.000000	0.065217	0.00	0.000000	
3	0.02	0.000000	0.000000	0.000000	0.060000	0.02	0.020000	
4	0.00	0.000000	0.000000	0.000000	0.040000	0.00	0.000000	
5	0.00	0.000000	0.000000	0.000000	0.040000	0.00	0.000000	
6	0.00	0.020000	0.000000	0.000000	0.040000	0.02	0.000000	
7	0.00	0.000000	0.000000	0.000000	0.102041	0.00	0.000000	
8	0.00	0.000000	0.000000	0.000000	0.120000	0.00	0.040000	
9	0.00	0.000000	0.000000	0.000000	0.040000	0.00	0.000000	
10	0.00	0.000000	0.000000	0.000000	0.060000	0.02	0.020000	
11	0.00	0.000000	0.000000	0.000000	0.080000	0.00	0.000000	
12	0.00	0.000000	0.000000	0.000000	0.040000	0.00	0.020000	
13	0.00	0.000000	0.000000	0.000000	0.140000	0.00	0.000000	
14	0.00	0.022222	0.022222	0.000000	0.022222	0.00	0.000000	
15	0.00	0.000000	0.000000	0.000000	0.102041	0.00	0.000000	
16	0.00	0.000000	0.000000	0.000000	0.100000	0.00	0.000000	
17	0.00	0.000000	0.000000	0.000000	0.040000	0.00	0.000000	
18	0.00	0.000000	0.000000	0.000000	0.060000	0.00	0.000000	
19	0.00	0.000000	0.000000	0.020833	0.125000	0.00	0.062500	
20	0.00	0.000000	0.022727	0.000000	0.090909	0.00	0.000000	
21	0.00	0.000000	0.000000	0.000000	0.022222	0.00	0.022222	
22	0.00	0.000000	0.000000	0.000000	0.076923	0.00	0.000000	
23	0.00	0.000000	0.000000	0.000000	0.000000	0.00	0.000000	
24	0.00	0.000000	0.000000	0.000000	0.040000	0.00	0.020000	
25	0.00	0.000000	0.000000	0.000000	0.095238	0.00	0.000000	
26	0.00	0.000000	0.020000	0.000000	0.020000	0.00	0.000000	
27	0.00	0.000000	0.000000	0.000000	0.140000	0.00	0.020000	
28	0.00	0.000000	0.000000	0.000000	0.083333	0.00	0.000000	

29	0.00	0.000000	0.000000	0.000000	0.040000	0.02	0.000000
30	0.00	0.000000	0.000000	0.000000	0.000000	0.00	0.000000
31	0.00	0.000000	0.000000	0.000000	0.060000	0.00	0.000000
32	0.00	0.000000	0.000000	0.000000	0.020000	0.04	0.000000
33	0.00	0.000000	0.000000	0.000000	0.040000	0.00	0.000000
34	0.00	0.000000	0.000000	0.000000	0.020000	0.00	0.040000
35	0.00	0.000000	0.000000	0.000000	0.040000	0.00	0.000000
36	0.00	0.000000	0.020000	0.000000	0.040000	0.00	0.000000
37	0.00	0.000000	0.000000	0.020000	0.100000	0.00	0.020000
38	0.02	0.000000	0.000000	0.000000	0.020000	0.02	0.000000
39	0.00	0.000000	0.020000	0.000000	0.120000	0.00	0.000000

	Restaurant	Russian	Salad	Sandwiches	Scandinavian	Seafood	\
0	0.000000	0.00	0.031250	0.062500	0.00	0.062500	
1	0.040000	0.00	0.020000	0.000000	0.00	0.000000	
2	0.021739	0.00	0.000000	0.043478	0.00	0.065217	
3	0.020000	0.00	0.000000	0.000000	0.00	0.020000	
4	0.000000	0.00	0.000000	0.000000	0.00	0.000000	
5	0.000000	0.00	0.000000	0.120000	0.00	0.000000	
6	0.040000	0.00	0.000000	0.040000	0.00	0.020000	
7	0.020408	0.00	0.000000	0.020408	0.00	0.020408	
8	0.000000	0.00	0.000000	0.000000	0.00	0.000000	
9	0.020000	0.02	0.060000	0.100000	0.00	0.020000	
10	0.020000	0.02	0.060000	0.060000	0.00	0.020000	
11	0.040000	0.00	0.000000	0.060000	0.00	0.000000	
12	0.000000	0.00	0.020000	0.020000	0.00	0.040000	
13	0.000000	0.00	0.000000	0.040000	0.00	0.020000	
14	0.066667	0.00	0.044444	0.044444	0.00	0.000000	
15	0.081633	0.00	0.000000	0.020408	0.00	0.020408	
16	0.060000	0.00	0.000000	0.000000	0.00	0.020000	
17	0.040000	0.00	0.000000	0.000000	0.00	0.020000	
18	0.040000	0.00	0.000000	0.000000	0.00	0.020000	
19	0.020833	0.00	0.000000	0.041667	0.00	0.000000	
20	0.000000	0.00	0.000000	0.000000	0.00	0.000000	
21	0.000000	0.00	0.000000	0.044444	0.00	0.066667	
22	0.000000	0.00	0.000000	0.230769	0.00	0.076923	
23	0.040000	0.00	0.000000	0.100000	0.00	0.000000	
24	0.060000	0.00	0.020000	0.060000	0.00	0.000000	
25	0.000000	0.00	0.023810	0.047619	0.00	0.023810	
26	0.020000	0.00	0.020000	0.080000	0.04	0.000000	
27	0.040000	0.00	0.000000	0.020000	0.00	0.000000	
28	0.083333	0.00	0.000000	0.166667	0.00	0.000000	
29	0.020000	0.00	0.020000	0.000000	0.00	0.040000	
30	0.000000	0.00	0.000000	0.200000	0.00	0.000000	
31	0.000000	0.00	0.000000	0.020000	0.00	0.000000	
32	0.020000	0.00	0.020000	0.040000	0.00	0.020000	
33	0.020000	0.00	0.020000	0.020000	0.00	0.000000	

34	0.020000	0.00	0.000000	0.060000	0.00	0.020000
35	0.000000	0.00	0.020000	0.040000	0.00	0.040000
36	0.020000	0.00	0.020000	0.000000	0.00	0.040000
37	0.060000	0.00	0.020000	0.000000	0.00	0.040000
38	0.000000	0.00	0.000000	0.020000	0.00	0.040000
39	0.000000	0.00	0.000000	0.080000	0.00	0.000000

	Shanghai	Snacks	Soup	South Indian	Southern / Soul	Spanish \
0	0.00	0.00	0.00	0.00	0.000000	0.000000
1	0.00	0.00	0.00	0.00	0.000000	0.000000
2	0.00	0.00	0.00	0.00	0.065217	0.000000
3	0.00	0.00	0.00	0.00	0.000000	0.020000
4	0.04	0.02	0.00	0.00	0.000000	0.000000
5	0.00	0.00	0.00	0.00	0.000000	0.000000
6	0.00	0.00	0.00	0.00	0.000000	0.000000
7	0.00	0.00	0.00	0.00	0.000000	0.040816
8	0.00	0.00	0.00	0.02	0.000000	0.000000
9	0.00	0.02	0.02	0.00	0.000000	0.000000
10	0.00	0.00	0.00	0.00	0.000000	0.020000
11	0.00	0.00	0.00	0.02	0.000000	0.000000
12	0.00	0.00	0.00	0.00	0.000000	0.000000
13	0.00	0.00	0.00	0.00	0.000000	0.020000
14	0.00	0.00	0.00	0.00	0.000000	0.066667
15	0.00	0.00	0.00	0.00	0.000000	0.040816
16	0.00	0.00	0.00	0.00	0.000000	0.000000
17	0.00	0.00	0.00	0.00	0.000000	0.000000
18	0.04	0.00	0.02	0.00	0.000000	0.020000
19	0.00	0.00	0.00	0.00	0.000000	0.000000
20	0.00	0.00	0.00	0.00	0.000000	0.000000
21	0.00	0.00	0.00	0.00	0.000000	0.022222
22	0.00	0.00	0.00	0.00	0.000000	0.000000
23	0.00	0.02	0.00	0.00	0.000000	0.000000
24	0.00	0.02	0.02	0.00	0.000000	0.000000
25	0.00	0.00	0.00	0.00	0.000000	0.000000
26	0.00	0.00	0.00	0.00	0.000000	0.000000
27	0.00	0.00	0.00	0.00	0.020000	0.000000
28	0.00	0.00	0.00	0.00	0.000000	0.000000
29	0.00	0.00	0.02	0.00	0.000000	0.020000
30	0.00	0.00	0.00	0.00	0.000000	0.000000
31	0.00	0.00	0.00	0.00	0.000000	0.000000
32	0.00	0.00	0.00	0.00	0.000000	0.020000
33	0.00	0.00	0.00	0.00	0.000000	0.020000
34	0.00	0.00	0.00	0.00	0.000000	0.020000
35	0.00	0.02	0.00	0.00	0.000000	0.000000
36	0.00	0.00	0.00	0.02	0.020000	0.000000
37	0.00	0.00	0.00	0.00	0.000000	0.040000
38	0.00	0.00	0.00	0.00	0.000000	0.000000

39	0.00	0.00	0.00	0.00	0.000000	0.000000
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	Steakhouse	Sushi	Szechuan	Tacos	Taiwanese	Tapas	Tex-Mex \
0	0.031250	0.031250	0.000000	0.000000	0.00	0.000000	0.000000
1	0.000000	0.060000	0.000000	0.040000	0.00	0.000000	0.000000
2	0.000000	0.000000	0.000000	0.000000	0.00	0.021739	0.021739
3	0.000000	0.060000	0.000000	0.000000	0.00	0.040000	0.000000
4	0.000000	0.000000	0.000000	0.020000	0.02	0.000000	0.000000
5	0.020000	0.060000	0.000000	0.020000	0.00	0.000000	0.000000
6	0.020000	0.000000	0.000000	0.040000	0.00	0.000000	0.000000
7	0.040816	0.000000	0.000000	0.020408	0.00	0.000000	0.000000
8	0.000000	0.020000	0.000000	0.020000	0.00	0.020000	0.000000
9	0.020000	0.000000	0.000000	0.000000	0.00	0.000000	0.000000
10	0.000000	0.000000	0.000000	0.020000	0.00	0.020000	0.000000
11	0.000000	0.020000	0.000000	0.040000	0.00	0.000000	0.000000
12	0.000000	0.080000	0.000000	0.000000	0.00	0.000000	0.000000
13	0.000000	0.040000	0.000000	0.000000	0.00	0.000000	0.000000
14	0.022222	0.000000	0.000000	0.000000	0.00	0.022222	0.000000
15	0.020408	0.000000	0.000000	0.020408	0.00	0.000000	0.000000
16	0.020000	0.020000	0.000000	0.000000	0.00	0.000000	0.000000
17	0.000000	0.000000	0.000000	0.020000	0.00	0.000000	0.000000
18	0.000000	0.000000	0.000000	0.000000	0.00	0.000000	0.000000
19	0.000000	0.000000	0.000000	0.000000	0.00	0.000000	0.000000
20	0.000000	0.022727	0.045455	0.000000	0.00	0.000000	0.000000
21	0.000000	0.044444	0.000000	0.000000	0.00	0.000000	0.000000
22	0.076923	0.000000	0.000000	0.000000	0.00	0.000000	0.000000
23	0.020000	0.040000	0.020000	0.000000	0.00	0.000000	0.000000
24	0.000000	0.000000	0.000000	0.000000	0.00	0.000000	0.000000
25	0.000000	0.000000	0.000000	0.000000	0.00	0.000000	0.000000
26	0.000000	0.020000	0.000000	0.000000	0.00	0.000000	0.000000
27	0.000000	0.020000	0.000000	0.000000	0.00	0.000000	0.000000
28	0.000000	0.000000	0.000000	0.000000	0.00	0.000000	0.000000
29	0.000000	0.000000	0.000000	0.000000	0.00	0.000000	0.000000
30	0.000000	0.200000	0.000000	0.000000	0.00	0.000000	0.000000
31	0.040000	0.020000	0.020000	0.000000	0.00	0.000000	0.000000
32	0.040000	0.020000	0.000000	0.000000	0.00	0.000000	0.000000
33	0.000000	0.060000	0.000000	0.000000	0.00	0.000000	0.000000
34	0.060000	0.040000	0.000000	0.020000	0.00	0.000000	0.000000
35	0.020000	0.060000	0.000000	0.000000	0.00	0.020000	0.000000
36	0.000000	0.020000	0.000000	0.000000	0.00	0.000000	0.000000
37	0.000000	0.020000	0.000000	0.000000	0.00	0.020000	0.000000
38	0.040000	0.040000	0.000000	0.020000	0.00	0.020000	0.000000
39	0.000000	0.060000	0.000000	0.020000	0.00	0.000000	0.000000

	Thai	Tonkatsu	Turkish	Udon	Vegetarian / Vegan	Vietnamese \
0	0.000000	0.00	0.00	0.00	0.000000	0.000000
1	0.020000	0.00	0.02	0.00	0.000000	0.000000

2	0.000000	0.00	0.00	0.00	0.021739	0.000000
3	0.020000	0.00	0.00	0.00	0.020000	0.020000
4	0.020000	0.00	0.00	0.00	0.060000	0.020000
5	0.000000	0.00	0.00	0.00	0.000000	0.000000
6	0.000000	0.00	0.00	0.00	0.020000	0.000000
7	0.061224	0.00	0.00	0.00	0.000000	0.000000
8	0.040000	0.00	0.00	0.02	0.060000	0.040000
9	0.000000	0.00	0.00	0.00	0.000000	0.000000
10	0.000000	0.00	0.02	0.00	0.000000	0.000000
11	0.100000	0.00	0.00	0.00	0.000000	0.040000
12	0.000000	0.00	0.00	0.00	0.040000	0.060000
13	0.000000	0.00	0.00	0.00	0.000000	0.000000
14	0.044444	0.00	0.00	0.00	0.022222	0.000000
15	0.040816	0.00	0.00	0.00	0.000000	0.000000
16	0.000000	0.00	0.02	0.00	0.020000	0.000000
17	0.000000	0.00	0.00	0.00	0.000000	0.000000
18	0.040000	0.00	0.02	0.00	0.020000	0.020000
19	0.020833	0.00	0.00	0.00	0.000000	0.020833
20	0.045455	0.00	0.00	0.00	0.000000	0.022727
21	0.000000	0.00	0.00	0.00	0.000000	0.000000
22	0.000000	0.00	0.00	0.00	0.000000	0.000000
23	0.000000	0.02	0.02	0.02	0.000000	0.020000
24	0.000000	0.00	0.00	0.00	0.020000	0.000000
25	0.000000	0.00	0.00	0.00	0.000000	0.000000
26	0.040000	0.00	0.02	0.00	0.040000	0.000000
27	0.000000	0.00	0.00	0.02	0.080000	0.020000
28	0.000000	0.00	0.00	0.00	0.000000	0.000000
29	0.020000	0.00	0.00	0.00	0.020000	0.000000
30	0.000000	0.00	0.00	0.00	0.000000	0.000000
31	0.040000	0.00	0.00	0.00	0.000000	0.020000
32	0.000000	0.00	0.00	0.00	0.000000	0.020000
33	0.040000	0.00	0.00	0.00	0.000000	0.000000
34	0.020000	0.00	0.04	0.00	0.000000	0.000000
35	0.020000	0.00	0.02	0.00	0.000000	0.000000
36	0.040000	0.00	0.02	0.00	0.040000	0.000000
37	0.000000	0.00	0.00	0.00	0.000000	0.000000
38	0.000000	0.00	0.00	0.00	0.040000	0.000000
39	0.020000	0.00	0.02	0.00	0.000000	0.020000

Wings

0	0.000000
1	0.000000
2	0.000000
3	0.000000
4	0.000000
5	0.020000
6	0.000000

```

7    0.000000
8    0.000000
9    0.000000
10   0.000000
11   0.000000
12   0.000000
13   0.000000
14   0.000000
15   0.000000
16   0.000000
17   0.000000
18   0.000000
19   0.000000
20   0.022727
21   0.000000
22   0.000000
23   0.000000
24   0.000000
25   0.000000
26   0.000000
27   0.000000
28   0.000000
29   0.000000
30   0.000000
31   0.000000
32   0.000000
33   0.020000
34   0.000000
35   0.000000
36   0.000000
37   0.000000
38   0.000000
39   0.020000

```

```

[40]: num_top_venues = 5

for hood in manhattan_grouped['Neighborhood']:
    print("----"+hood+"----")
    temp = manhattan_grouped[manhattan_grouped['Neighborhood'] == hood].T.
    →reset_index()
    temp.columns = ['venue', 'freq']
    temp = temp.iloc[1:]
    temp['freq'] = temp['freq'].astype(float)
    temp = temp.round({'freq': 2})
    print(temp.sort_values('freq', ascending=False).reset_index(drop=True).
    →head(num_top_venues))
    print('\n')

```

----Battery Park City----

	venue	freq
0	Pizza	0.12
1	Italian	0.12
2	Chinese	0.09
3	Food Court	0.06
4	Burgers	0.06

----Carnegie Hill----

	venue	freq
0	Bakery	0.12
1	Pizza	0.12
2	Café	0.10
3	French	0.06
4	Sushi	0.06

----Central Harlem----

	venue	freq
0	Deli / Bodega	0.11
1	Fried Chicken	0.09
2	Chinese	0.09
3	African	0.07
4	Pizza	0.07

----Chelsea----

	venue	freq
0	Bakery	0.10
1	Italian	0.10
2	Café	0.08
3	Sushi	0.06
4	American	0.06

----Chinatown----

	venue	freq
0	Chinese	0.24
1	Dumplings	0.08
2	Vegetarian / Vegan	0.06
3	Italian	0.06
4	Bakery	0.06

----Civic Center----

	venue	freq
0	Sandwiches	0.12

1	Italian	0.10
2	French	0.08
3	American	0.08
4	Bakery	0.06

----Clinton----

	venue	freq
0	Deli / Bodega	0.16
1	Italian	0.10
2	Café	0.06
3	Mexican	0.04
4	French	0.04

----East Harlem----

	venue	freq
0	Mexican	0.14
1	Deli / Bodega	0.10
2	Pizza	0.10
3	Bakery	0.10
4	Latin American	0.08

----East Village----

	venue	freq
0	Pizza	0.12
1	Japanese	0.10
2	Italian	0.08
3	Vegetarian / Vegan	0.06
4	American	0.06

----Financial District----

	venue	freq
0	Sandwiches	0.10
1	Café	0.08
2	American	0.08
3	Food Truck	0.06
4	Salad	0.06

----Flatiron----

	venue	freq
0	Italian	0.16
1	Salad	0.06
2	Korean	0.06
3	Sandwiches	0.06

4 Pizza 0.06

----Gramercy----

	venue	freq
0	Deli / Bodega	0.10
1	Italian	0.10
2	Thai	0.10
3	Pizza	0.08
4	Mexican	0.08

----Greenwich Village----

	venue	freq
0	Italian	0.18
1	French	0.12
2	Sushi	0.08
3	Café	0.08
4	Vietnamese	0.06

----Hamilton Heights----

	venue	freq
0	Pizza	0.14
1	Deli / Bodega	0.14
2	Café	0.08
3	Chinese	0.08
4	Mexican	0.08

----Hudson Yards----

	venue	freq
0	American	0.11
1	Café	0.09
2	Italian	0.09
3	Spanish	0.07
4	Restaurant	0.07

----Inwood----

	venue	freq
0	Pizza	0.10
1	Restaurant	0.08
2	Mexican	0.08
3	Café	0.08
4	Bakery	0.06

----Lenox Hill----

	venue	freq
0	Italian	0.16
1	Pizza	0.10
2	Deli / Bodega	0.10
3	Restaurant	0.06
4	Greek	0.06

----Lincoln Square----

	venue	freq
0	Café	0.16
1	Food Truck	0.14
2	Italian	0.10
3	American	0.10
4	French	0.06

----Little Italy----

	venue	freq
0	Italian	0.14
1	Bakery	0.10
2	Chinese	0.08
3	Café	0.08
4	Pizza	0.06

----Lower East Side----

	venue	freq
0	Pizza	0.12
1	Deli / Bodega	0.10
2	Chinese	0.08
3	Bakery	0.06
4	Ramen	0.06

----Manhattan Valley----

	venue	freq
0	Mexican	0.11
1	Pizza	0.09
2	Indian	0.07
3	Szechuan	0.05
4	Chinese	0.05

----Manhattanville----

	venue	freq
0	Chinese	0.11

1	Deli / Bodega	0.09
2	Mexican	0.07
3	Seafood	0.07
4	Bakery	0.04

----Marble Hill----

	venue	freq
0	Sandwiches	0.23
1	Deli / Bodega	0.15
2	Donuts	0.08
3	Pizza	0.08
4	Seafood	0.08

----Midtown----

	venue	freq
0	Sandwiches	0.10
1	Food Truck	0.08
2	Deli / Bodega	0.08
3	American	0.06
4	Burgers	0.04

----Midtown South----

	venue	freq
0	Korean	0.16
1	Japanese	0.12
2	Bakery	0.10
3	Deli / Bodega	0.08
4	Sandwiches	0.06

----Morningside Heights----

	venue	freq
0	Food Truck	0.14
1	Deli / Bodega	0.12
2	Pizza	0.10
3	American	0.07
4	Café	0.07

----Murray Hill----

	venue	freq
0	American	0.12
1	Sandwiches	0.08
2	Scandinavian	0.04
3	Thai	0.04

4 Mexican 0.04

----Noho----

	venue	freq
0	Italian	0.16
1	Pizza	0.14
2	Vegetarian / Vegan	0.08
3	Japanese	0.08
4	American	0.04

----Roosevelt Island----

	venue	freq
0	Sandwiches	0.17
1	Deli / Bodega	0.17
2	Japanese	0.08
3	Café	0.08
4	Kosher	0.08

----Soho----

	venue	freq
0	Italian	0.16
1	Café	0.12
2	French	0.12
3	Bakery	0.08
4	Mediterranean	0.06

----Stuyvesant Town----

	venue	freq
0	Sushi	0.2
1	Sandwiches	0.2
2	Deli / Bodega	0.2
3	Bistro	0.2
4	German	0.2

----Sutton Place----

	venue	freq
0	American	0.10
1	Bagels	0.10
2	Italian	0.08
3	Pizza	0.06
4	Chinese	0.06

----Tribeca----

	venue	freq
0	Italian	0.10
1	American	0.10
2	Bakery	0.06
3	Café	0.06
4	Deli / Bodega	0.06

----Tudor City----

	venue	freq
0	Café	0.12
1	Deli / Bodega	0.12
2	Mexican	0.10
3	Sushi	0.06
4	American	0.06

----Turtle Bay----

	venue	freq
0	Café	0.12
1	Italian	0.12
2	Food Truck	0.08
3	Deli / Bodega	0.08
4	Sandwiches	0.06

----Upper East Side----

	venue	freq
0	Italian	0.20
1	American	0.08
2	Diner	0.06
3	French	0.06
4	Sushi	0.06

----Upper West Side----

	venue	freq
0	Italian	0.14
1	Bakery	0.06
2	French	0.04
3	American	0.04
4	Café	0.04

----Washington Heights----

	venue	freq
0	Deli / Bodega	0.18

1	Pizza	0.10
2	Latin American	0.06
3	Restaurant	0.06
4	Chinese	0.06

----West Village----

	venue	freq
0	Italian	0.14
1	Gastropub	0.06
2	American	0.06
3	Deli / Bodega	0.04
4	Sushi	0.04

----Yorkville----

	venue	freq
0	Italian	0.12
1	Pizza	0.12
2	Deli / Bodega	0.10
3	Sandwiches	0.08
4	Chinese	0.06

```
[41]: def return_most_common_venues(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]
```

```
[81]: 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('{}{}'.format(ind+1, indicators[ind]))
          except:
              columns.append('{}th'.format(ind+1))

      # create a new dataframe
      neighborhoods_venues_sorted = pd.DataFrame(columns=columns)
      neighborhoods_venues_sorted['Neighborhood'] = manhattan_grouped['Neighborhood']

      for ind in np.arange(manhattan_grouped.shape[0]):
```

```
neighborhoods_venues_sorted.iloc[ind, 1:] =
→return_most_common_venues(manhattan_grouped.iloc[ind, :], num_top_venues)

neighborhoods_venues_sorted.head()
```

```
[81]:
```

	Neighborhood	1st	2nd	3rd \
0	Battery Park City	Pizza	Italian	Chinese
1	Carnegie Hill	Bakery	Pizza	Café
2	Central Harlem	Deli / Bodega	Fried Chicken	Chinese
3	Chelsea	Italian	Bakery	Café
4	Chinatown	Chinese	Dumplings	Malay

	4th	5th
0	Sandwiches	BBQ
1	French	Sushi
2	African	Southern / Soul
3	Pizza	Sushi
4	Vegetarian / Vegan	Italian

```
[45]: with open('neighborhoods_venues.tex', 'w') as f:
        f.write(neighborhoods_venues_sorted.head().to_latex(index=False))
```

```
[82]: # set number of clusters
kclusters = 4

manhattan_grouped_clustering = manhattan_grouped.drop('Neighborhood', 1)

# run k-means clustering
kmeans = KMeans(n_clusters=kclusters, random_state=0).
→fit(manhattan_grouped_clustering)

# check cluster labels generated for each row in the dataframe
kmeans.labels_[0:10]
```

```
[82]: array([0, 1, 1, 0, 1, 2, 2, 1, 1, 2], dtype=int32)
```

```
[83]: # add clustering labels
neighborhoods_venues_sorted.insert(0, 'Cluster Labels', kmeans.
→predict(manhattan_grouped_clustering))

manhattan_merged = manhattan_data

# merge toronto_grouped with toronto_data to add latitude/longitude for each
→neighborhood
manhattan_merged = manhattan_merged.join(neighborhoods_venues_sorted.
→set_index('Neighborhood'), on='Neighborhood')

manhattan_merged.head() # check the last columns!
```

```
[83]:
```

	Borough	Neighborhood	Latitude	Longitude	Cluster	Labels \
0	Manhattan	Marble Hill	40.876551	-73.910660	1	
1	Manhattan	Chinatown	40.715618	-73.994279	1	
2	Manhattan	Washington Heights	40.851903	-73.936900	1	
3	Manhattan	Inwood	40.867684	-73.921210	1	
4	Manhattan	Hamilton Heights	40.823604	-73.949688	1	

	1st	2nd	3rd	4th	5th
0	Sandwiches	Deli / Bodega	Donuts	Seafood	Diner
1	Chinese	Dumplings	Malay	Vegetarian / Vegan	Italian
2	Deli / Bodega	Pizza	Mexican	Chinese	Restaurant
3	Pizza	Café	Restaurant	Mexican	Bakery
4	Pizza	Deli / Bodega	Café	Mexican	Chinese

```
[84]: # create map
map_clusters = folium.Map(location=[latitude, longitude], zoom_start=11)

# 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(manhattan_merged['Latitude'],
    ↳manhattan_merged['Longitude'], manhattan_merged['Neighborhood'],
    ↳manhattan_merged['Cluster Labels']):
    label = folium.Popup(str(poi) + ' Cluster ' + str(cluster), parse_html=True)
    folium.CircleMarker(
        [lat, lon],
        radius=5,
        popup=label,
        color=rainbow[cluster-1],
        fill=True,
        fill_color=rainbow[cluster-1],
        fill_opacity=0.7).add_to(map_clusters)

map_clusters
```

```
[84]: <folium.folium.Map at 0x7f6d4f7e4400>
```

```
[85]: manhattan_merged.loc[manhattan_merged['Cluster Labels'] == 0, manhattan_merged.
    ↳columns[[1] + list(range(5, manhattan_merged.shape[1]))]].head()
```

```
[85]:
```

	Neighborhood	1st	2nd	3rd	4th	5th
8	Upper East Side	Italian	American	French	Burgers	Sushi
10	Lenox Hill	Italian	Deli / Bodega	Pizza	Greek	Restaurant
12	Upper West Side	Italian	Bakery	Café	Mexican	French

17	Chelsea	Italian	Bakery	Café	Pizza	Sushi
18	Greenwich Village	Italian	French	Café	Sushi	Vietnamese

```
[86]: manhattan_merged.loc[manhattan_merged['Cluster Labels'] == 1, manhattan_merged.
      ↪columns[[1] + list(range(5, manhattan_merged.shape[1]))]]
```

```
[86]:
```

	Neighborhood	1st	2nd	3rd \
0	Marble Hill	Sandwiches	Deli / Bodega	Donuts
1	Chinatown	Chinese	Dumplings	Malay
2	Washington Heights	Deli / Bodega	Pizza	Mexican
3	Inwood	Pizza	Café	Restaurant
4	Hamilton Heights	Pizza	Deli / Bodega	Café
5	Manhattanville	Chinese	Deli / Bodega	Seafood
6	Central Harlem	Deli / Bodega	Fried Chicken	Chinese
7	East Harlem	Mexican	Pizza	Deli / Bodega
9	Yorkville	Italian	Pizza	Deli / Bodega
11	Roosevelt Island	Deli / Bodega	Sandwiches	Café
19	East Village	Pizza	Japanese	Italian
20	Lower East Side	Pizza	Deli / Bodega	Chinese
25	Manhattan Valley	Mexican	Pizza	Indian
26	Morningside Heights	Food Truck	Deli / Bodega	Pizza
27	Gramercy	Deli / Bodega	Thai	Italian
30	Carnegie Hill	Bakery	Pizza	Café
33	Midtown South	Korean	Japanese	Bakery
36	Tudor City	Café	Deli / Bodega	Mexican

	4th	5th
0	Seafood	Diner
1	Vegetarian / Vegan	Italian
2	Chinese	Restaurant
3	Mexican	Bakery
4	Mexican	Chinese
5	Mexican	Sandwiches
6	African	Southern / Soul
7	Bakery	Latin American
9	Sandwiches	Chinese
11	Pizza	Greek
19	Vegetarian / Vegan	American
20	Mexican	Ramen
25	Thai	Café
26	American	Café
27	Pizza	Mexican
30	French	Sushi
33	Deli / Bodega	Restaurant
36	American	Food Truck

```
[90]: with open('mexican_cluster.tex', 'w') as f:
```

```
f.write(manhattan_merged.loc[manhattan_merged['Cluster Labels'] == 1,
↳manhattan_merged.columns[[1] + list(range(5, manhattan_merged.shape[1]))]].
↳to_latex(index=False))
```

```
[87]: manhattan_merged.loc[manhattan_merged['Cluster Labels'] == 2, manhattan_merged.
↳columns[[1] + list(range(5, manhattan_merged.shape[1]))]]
```

```
[87]:
```

	Neighborhood	1st	2nd	3rd	4th \
13	Lincoln Square	Café	Food Truck	American	Italian
14	Clinton	Deli / Bodega	Italian	Café	Sandwiches
15	Midtown	Sandwiches	Food Truck	Deli / Bodega	American
16	Murray Hill	American	Sandwiches	Chinese	Japanese
21	Tribeca	Italian	American	Bakery	Café
24	West Village	Italian	American	Gastropub	Burgers
29	Financial District	Sandwiches	Café	American	Salad
32	Civic Center	Sandwiches	Italian	American	French
34	Sutton Place	Bagels	American	Italian	Chinese
35	Turtle Bay	Italian	Café	Deli / Bodega	Food Truck
39	Hudson Yards	American	Café	Italian	Spanish

	5th
13	French
14	Restaurant
15	Burgers
16	Scandinavian
21	Burgers
24	Steakhouse
29	Food Truck
32	Bakery
34	Pizza
35	Sandwiches
39	Restaurant

```
[88]: manhattan_merged.loc[manhattan_merged['Cluster Labels'] == 3, manhattan_merged.
↳columns[[1] + list(range(5, manhattan_merged.shape[1]))]]
```

```
[88]:
```

	Neighborhood	1st	2nd	3rd	4th	5th
37	Stuyvesant Town	Deli / Bodega	Sandwiches	German	Bistro	Sushi