Capstone Project - The Battle of Neighborhoods: Week 4 Assignment

Title: Comparing Neighborhoods in LA to SF and NY

1- Problem Description and Background Discussion

A producer of deli meat located in California supplies various types of cured meat to restaurants, supermarkets and sandwich shops located in the Los Angeles metropolitan area. His business is thriving and is considering expanding to another populous city with a similar profile of business distribution and people preferences. Since his business model takes advantage of the appetite for cured meats in Mexican and Italian restaurants and would benefit from expanding to a city with a similar distribution of venues.

To help this producer decide between San Francisco and New York City, we will carry out an analysis of the type of venues and their distribution in each city and determine which of the two shares more similarity with Los Angeles, hence presenting a better chance of success.

We will first collect data on the neighborhoods of all three cities, including name, zip code and geographical coordinates. Then we will extract information on the venues located within each neighborhood. A clustering of the neighborhoods based on the type of venues they contain will follow, together with analysis of the types of venues and plots of the cluster distribution on the map. Finally, we will estimate the similarity between cities to provide a recommendation.

2- Data Description and Application for Solving the Problem

The extraction and processing of the source data followed several steps:

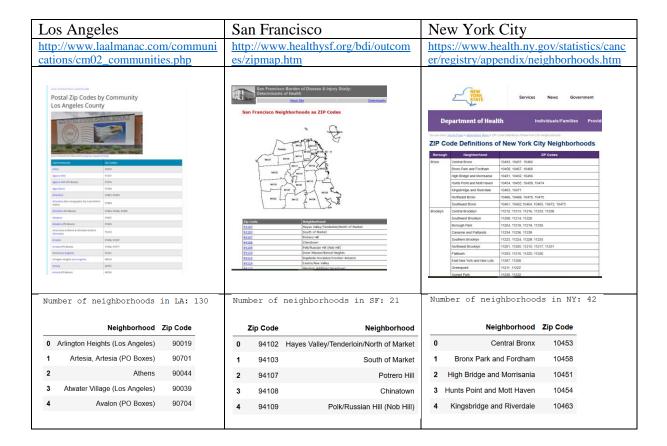
a) Web Data Scraping:

To obtain geographical information from each city, one web site was selected in each case that included a table with at least zip-codes and neighborhood names for that city.

The data was scraped using the library **requests** to grab html data and the library **BeautifulSoup** to scrape html data.

The data for each city was wrangled separately according to the characteristics of the URL source and converted to a dataframe with the columns 'Neighborhood' and 'Zip Code'.

The table below collects the URL information, an image of the site and the resulting dataframe for each of the three cities:



b) Coordinates Extraction:

The coordinates of each neighborhood with zip code was extracted using the function **Nominatim** from the library **geopy.geocoders**, which converts an address into latitude and longitude values. For simplicity and to reduce ambiguity given that some zip codes included more than one neighborhood name, the zip code value was used to obtain the latitude and longitude values. Whenever a zip code did not return valid latitude or longitude coordinates, the entire row was deleted. The data was then added to the dataframe of each corresponding city under the columns 'Latitude' and 'Longitude'.

The table below displays the first few rows of each dataframe with all the necessary geographical information to extract venue information in the next step.

Lo	s Angeles				Sar	n Fra	ıncisco	New York City							
Number of neighborhoods in LA: 125					Num	ber of	neighborhoods in SF: 21	Number of neighborhoods in NY: 42							
Neighborhood		Zip Code Latitud		Longitude	3	Zip Code	Neighborhood	Latitude	Longitude			Neighborhood		Latitude	Longitude
0	Arlington Heights (Los Angeles)	90019	34.047371	-118.336046	0	94102	Hayes Valley/Tenderloin/North of Market	37.779491	-122.418224	0		Central Bronx	10453	40.852348	-73.911965
1	Artesia, Artesia (PO Boxes)	90701	33.868528	-118.077698	1	94103	South of Market	37.774425	-122.411091	1	Bronx P	ark and Fordham	10458	40.861569	-73.888765
2	Athens	90044	33.981914	-118.287489	2	94107	Potrero Hill	37.793634	-122.408295	2	High Bride	ge and Morrisania	10451	40 828384	-73.927084
3	Atwater Village (Los Angeles)	90039	34.118121	-118.264129	3	94108	Chinatown	37.791043	-122.406578						
4	Avalon (PO Boxes)	90704	33.341730	-118.328136	4	94109	Polk/Russian Hill (Nob Hill)	37.793815	-122.420597	4		it and Mott Haven ige and Riverdale			-73.918198 -73.887248

c) Venue Extraction:

With the coordinates of each neighborhood of each city collected into each corresponding dataframe, the **FourSquare API** set up in week 1 of the course was employed to extract common venues within a pre-defined radius per neighborhood. Because the neighborhoods in Los Angeles can extend a larger area than those in San Francisco or New York, the search radius was adjusted slightly for each city. In particular, a radius of 1000 meters was selected for Los Angeles while 500 meters was selected for both San Francisco and New York City, respectively.

Using code from previous weeks, only relevant information about each returned venue was collected into three separate dataframes, one per each city.

The table below displays the first few rows of the venue dataframes extracted from FourSquare:

		Venue V	enue La	titude	Venue	Long	itude	Venu	ie Categor	,
)	Piz	zaRev	34.04	48585	-1	18.33	6439		Pizza Place	-
	Smart & Final	Extra!	34.04	47692	-1	18.33	5932	G	rocery Store	•
2	Planet F	itness	34.04	47774	-1	18.33	8605	Gym / Fitr	ess Cente	r
;	Jersey Mike's	Subs	34.04	48449	-1	18.33	7419	San	dwich Place	•
ļ	Pei	Smart	34.04	48184	-1	18.33	5489		Pet Store	•
5	La Fayette S	quare	34.04	43205	-1	18.33	3813	Ne	ighborhoo	i
i	Midtown Cro	ssing	34.04	48047	-1	18.33	7077	Sh	opping Mal	ı
1	Mateo's Ice Cream & Fru	it Bars	34.04	47588	-1	18.32	7972	Ice C	ream Sho	
1	El Co	mpita	34.04	48592	-1	18.33	2846	Mexican	Restauran	t
•	Panda Express M	id-City	34.04	48654	-1	18.33	7556	Chinese	Restauran	t
٦,	n Francisco									
aı	TTTATICISCO									
_				enue La			e Long			enue Category
0		n Art Muse			30178		122.41			Art Museum
1	Louise M. Davies				77976		122.42			Concert Hall
2	,	Herbst The Philz Co			79548		122.42 122.41			Coffee Shop
4	War Memorial				78601		122.41			Opera House
5		ancisco B			78580		122.42			Dance Studio
6								.416353 Vegetarian / Veg		
7	Siam Orchid Traditional				77111		122.41			lassage Studio
8		August 1	-		30537		122.42			tian Restaurant
9	Wark	lemorial C			79042		122.42			Park
le	w York City									
	Venue	Venue La	titude	Venue	Longi	tude		Venu	ie Categor	у
0	Liberato	40.8	53744		-73.907	7966	Latin	American	Restaura	nt
	Accra Resturant	40.8	53871		-73.90	3421		African	Restaura	nt
1						7200			Wings Joir	nt
2	Wingstop	40.8	54093		-73.907	000				
	Wingstop Bravo Supermarkets		54093 53936		-73.90 -73.91			G	rocery Stor	е
2		40.8				1144		G	rocery Stor Pizza Plac	
2	Bravo Supermarkets	40.8 40.8	53936		-73.91	1144 8976				е
2 3 4	Bravo Supermarkets Papa John's Pizza Dunkin Donuts	40.8 40.8	53936 52429		-73.914 -73.900 -73.900	1144 3976 3724			Pizza Plac	e p
2 3 4 5	Bravo Supermarkets Papa John's Pizza Dunkin Donuts Chase Bank	40.8 40.8 40.8	53936 52429 53817 54087		-73.914 -73.906 -73.906 -73.907	3976 3724 7631			Pizza Plac Donut Sho Ban	e p k
2 3 4 5	Bravo Supermarkets Papa John's Pizza Dunkin Donuts	40.8 40.8 40.8 40.8	53936 52429 53817		-73.914 -73.900 -73.900	1144 3976 3724 7631 9267		s	Pizza Plac Donut Sho	e p k

These dataframes are used in subsequent steps to analyze the neighborhood information and extract insights.