WEEK 5: FINAL CAPSTONE PROJECT THE BATTLE OF NEIGHBORHOODS

INTRODUCTION

- For this Capstone Week 5 project, I am creating a scenario which can help New York government to open new 'Pharmacy Stores' across neighbourhoods of New York City. The idea behind this project is that there may not be enough Pharmacy stores and it might present a great help to the Government and Health department.
- My purpose in mind is of finding the locations across New York which lack in Pharmacy stores.



BUSINESS PROBLEM

- The objective of this capstone project is to find all the suitable locations where Pharmacy stores can be built. By using all the skills and techniques of machine learning and data science, this project aims at providing solutions to following business question: In which locations should Government and Health department focus in opening pharmacy stores?
- There are 5 Boroughs in New York City namely Bronx, Brooklyn, Manhattan, Queens and Staten Island.
- After this project we will be able to list out all the neighbourhoods of New York City which lack in Pharmacy stores in two groups:
- Neighbourhoods of Brooklyn and Manhattan where there is shortage of Pharmacy Stores.
- Neighbourhoods of Bronx, Queens and Staten Island where there is shortage of Pharmacy Stores.



DATA SECTION:

- City to be analysed in this project: New York
- The New York City has 5 boroughs and 306 neighbourhoods. In order to segment and explore them, we will essentially need a dataset contains the 5 boroughs and the neighbourhoods that exist in each borough as well as the latitude and longitude coordinates of each neighbourhood. The following dataset contains all this information:

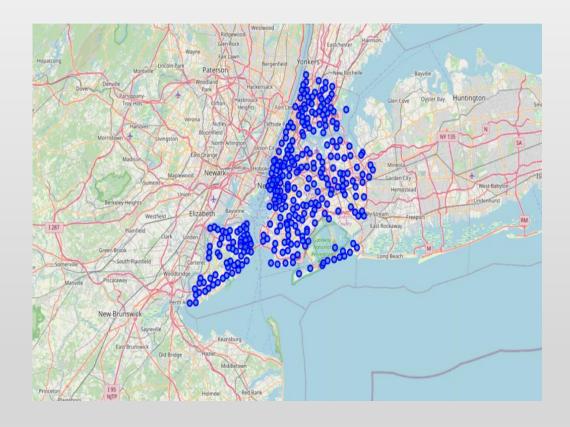
https://cocl.us/new_york_dataset

• By using Foursquare API we will get all the information about pharmacy stores in each neighbourhood of New York. By using this API we will get all the venues of neighbourhood then we have to filter these venues to get only Pharmacy stores.

METHODOLOGY

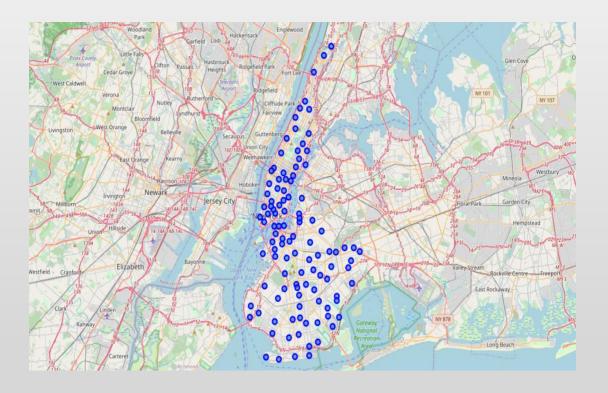
New York City neighbourhood has a total of 5 boroughs and 306 neighbourhoods.

[9]:					
ouc[s].		Borough	Neighborhood	Latitude	Longitude
	0	Bronx	Wakefield	40.894705	-73.847201
	1	Bronx	Co-op City	40.874294	-73.829939
	2	Bronx	Eastchester	40.887556	-73.827806
	3	Bronx	Fieldston	40.895437	-73.905643
	4	Bronx	Riverdale	40.890834	-73.912585
	5	Bronx	Kingsbridge	40.881687	-73.902818
	6	Manhattan	Marble Hill	40.876551	-73.910660
	7	Bronx	Woodlawn	40.898273	-73.867315
	8	Bronx	Norwood	40.877224	-73.879391
	9	Bronx	Williamsbridge	40.881039	-73.857446
	10	Bronx	Baychester	40.866858	-73.835798
	11	Bronx	Pelham Parkway	40.857413	-73.854756
	12	Bronx	City Island	40.847247	-73.786488
	13	Bronx	Bedford Park	40.870185	-73.885512
	14	Bronx	University Heights	40.855727	-73.910416



Separating the neighbourhoods and their data of Brooklyn and Manhattan from the original data to perform clustering and listing the possible neighbourhoods for our project.

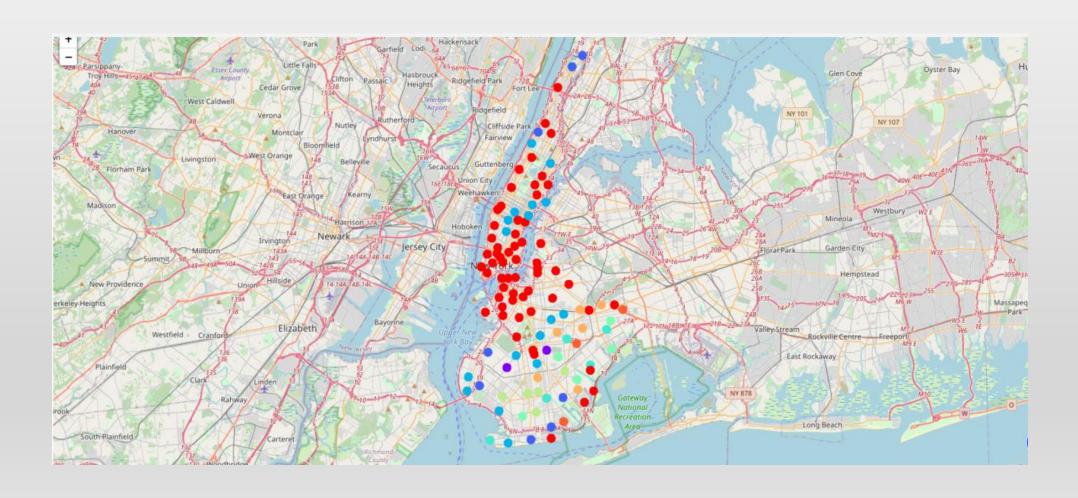
bm_c	data	a.head(15)		
		Borough	Neighborhood	Latitude	Longitude
	0	Manhattan	Marble Hill	40.876551	-73.910660
	1	Brooklyn	Bay Ridge	40.625801	-74.030621
	2	Brooklyn	Bensonhurst	40.611009	-73.995180
	3	Brooklyn	Sunset Park	40.645103	-74.010316
	4	Brooklyn	Greenpoint	40.730201	-73.954241
	5	Brooklyn	Gravesend	40.595260	-73.973471
	6	Brooklyn	Brighton Beach	40.576825	-73.965094
	7	Brooklyn	Sheepshead Bay	40.586890	-73.943186
	8	Brooklyn	Manhattan Terrace	40.614433	-73.957438
	9	Brooklyn	Flatbush	40.636326	-73.958401
	10	Brooklyn	Crown Heights	40.670829	-73.943291
	11	Brooklyn	East Flatbush	40.641718	-73.936103
	12	Brooklyn	Kensington	40.642382	-73.980421
	13	Brooklyn	Windsor Terrace	40.656946	-73.980073
	14	Brooklyn	Prospect Heights	40.676822	-73.964859



Using foursquare API we list out all the venues and category of venues of neighbourhoods of Brooklyn and Manhattan.

[25]: bm_	venue	es.head(20)						
Out[25]:	ı	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category
	0	Marble Hill	40.876551	-73.91066	Bikram Yoga	40.876844	-73.906204	Yoga Studio
	1	Marble Hill	40.876551	-73.91066	Arturo's	40.874412	-73.910271	Pizza Place
	2	Marble Hill	40.876551	-73.91066	Tibbett Diner	40.880404	-73.908937	Diner
	3	Marble Hill	40.876551	-73.91066	Sam's Pizza	40.879435	-73.905859	Pizza Place
	4	Marble Hill	40.876551	-73.91066	Starbucks	40.877531	-73.905582	Coffee Shop
	5	Marble Hill	40.876551	-73.91066	El Malecon	40.879338	-73.904457	Caribbean Restaurant
	6	Marble Hill	40.876551	-73.91066	Baker Athletic Complex	40.872061	-73.914876	Athletics & Sports
	7	Marble Hill	40.876551	-73.91066	The Bronx Public	40.878377	-73.903481	Pub
	8	Marble Hill	40.876551	-73.91066	Estrellita Poblana V	40.879687	-73.906257	Mexican Restaurant

Following map we can see the different types of clusters created by using k-means for Brooklyn and Manhattan by filtering data and listing only Pharmacy Stores.

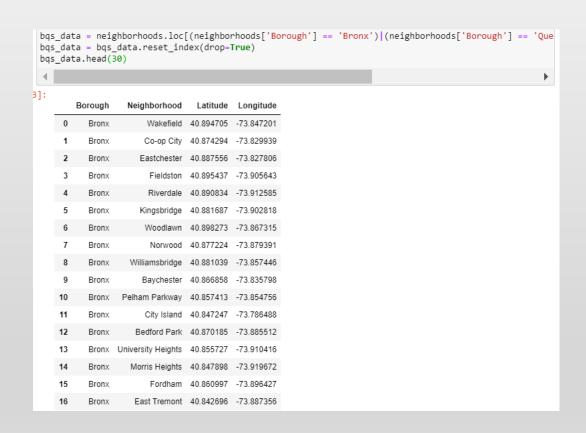


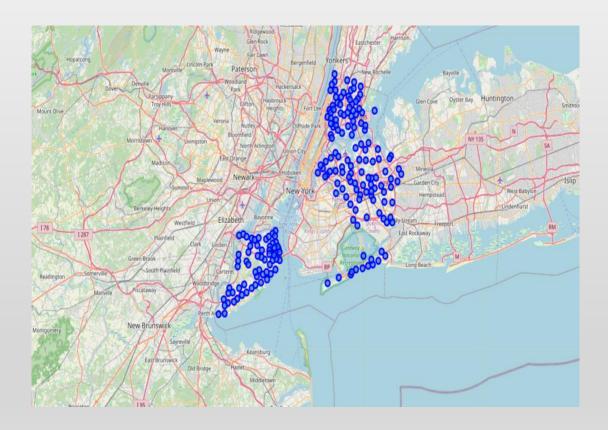
RESULT FOR PART 1 (BROOKLYN & MANHATTAN)

Through clustering we get that cluster – 0 contains the list of all neighbourhoods where there is shortage of Pharmacy Stores.

Cluster 0										
#Clu clus clus	ter	0 = bm_merge	d.loc[bm_r	nerged['Cluster Labe	els'] == 0]				
]:		Neighborhood	Pharmacy	Cluster Labels	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	
	1	Battery Park City	0.0	0	40.711932	-74.016869	Battery Park City Esplanade	40.711622	-74.017907	
	1	Battery Park City	0.0	0	40.711932	-74.016869	Institute of Culinary Education	40.712399	-74.015971	
	1	Battery Park City	0.0	0	40.711932	-74.016869	Hudson Eats	40.712666	-74.015901	
	1	Battery Park City	0.0	0	40.711932	-74.016869	Waterfront Plaza, Brookfield Place	40.713241	-74.016241	
	1	Battery Park City	0.0	0	40.711932	-74.016869	Equinox Brookfield Place	40.712704	-74.014995	
	1	Battery Park City	0.0	0	40.711932	-74.016869	Brookfield Place (BFPL)	40.713240	-74.015193	
							National Contember 44		•	

Separating the neighbourhoods and their data of Bronx, Queens and Staten Island from the original data to perform clustering and listing the possible neighbourhoods for our project.

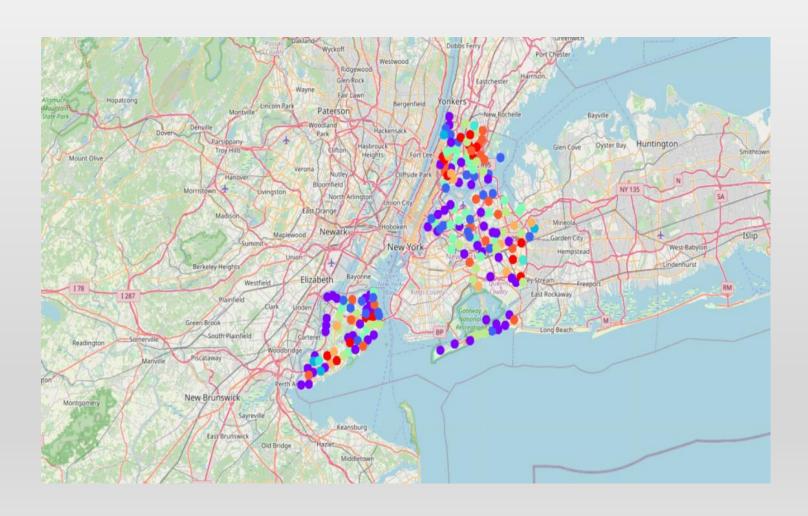




Using foursquare API we list out all the venues and category of venues of neighbourhoods of Bronx, Queens and Staten Island.

_venu	es.head(25)						
N	eighborhood Neighborhood Latitude Longitude		Venue	Venue Latitude	Venue Longitude	Venue Catego	
0	Wakefield	40.894705	-73.847201	Lollipops Gelato	40.894123	-73.845892	Dessert Sho
1	Wakefield	40.894705	-73.847201	Ripe Kitchen & Bar	40.898152	-73.838875	Caribbea Restaurar
2	Wakefield	40.894705	-73.847201	Ali's Roti Shop	40.894036	-73.856935	Caribbea Restaura
3	Wakefield	40.894705	-73.847201	Carvel Ice Cream	40.890487	-73.848568	Ice Cream Sho
4	Wakefield	40.894705	-73.847201	Jackie's West Indian Bakery	40.889283	-73.843310	Caribbea Restaura
5	Wakefield	40.894705	-73.847201	Jimbo's	40.891740	-73.858226	Burger Joi
6	Wakefield	40.894705	-73.847201	Dunkin'	40.890459	-73.849089	Donut Sho
7	Wakefield	40.894705	-73.847201	Rite Aid	40.889062	-73.842993	Pharmad
8	Wakefield	40.894705	-73.847201	Walgreens	40.896528	-73.844700	Pharma
9	Wakefield	40.894705	-73.847201	Rite Aid	40.896649	-73.844846	Pharma
10	Wakefield	40.894705	-73.847201	Subway	40.890468	-73.849152	Sandwich Place
11	Wakefield	40.894705	-73.847201	Shell	40.894187	-73.845862	Gas Statio
12	Wakefield	40.894705	-73.847201	E&L Bakery	40.893564	-73.856997	Bake

Following map we can see the different types of clusters created by using k-means for Bronx, Queens and Staten Island by filtering data and listing only Pharmacy Stores.



RESULT FOR PART 2 (BRONX, QUEENS AND STATEN ISLAND)

Through clustering we get that cluster – 1 contains the list of all neighbourhoods where there is shortage of Pharmacy Stores.

	CI	uster 1								
c.	<i>Cluste</i> luste luste	r1 = bqs_merg	ed.loc[bq:	s_merge	d['Cluster La	abels'] == 1]				
[80]:		Neighborhood	Pharmacy	Cluster Labels	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	
	3	Arlington	0.0	1	40.635325	-74.165104	Kohl's	40.626935	-74.164104	D€
	3	Arlington	0.0	1	40.635325	-74.165104	Comic Book Jones	40.626547	-74.163384	Со
	3	Arlington	0.0	1	40.635325	-74.165104	The Home Depot	40.628964	-74.172358	ı
	3	Arlington	0.0	1	40.635325	-74.165104	7-Eleven	40.626979	-74.165870	Cor
	3	Arlington	0.0	1	40.635325	-74.165104	Lowe's	40.627646	-74.162136	ı
	3	Arlington	0.0	1	40.635325	-74.165104	P And E Custom Kitchen Inc	40.634743	-74.165784	Hom
							MTA Due South			+

CONCLUSION - 1

For Manhattan and Brooklyn

The neighbourhoods are:

Battery Park City, Bedford Stuyvesant, Boerum Hill, Bergen Beach, Broadway Junction, Brooklyn Heights, Bushwick, Carnegie Hill, Carroll Gardens, Central Harlem, Chelsea, Chinatown, Civic Center, Clinton, Clinton Hill, Cobble Hill, Cypress Hills, Ditmas Park, Downtown, Dumbo, East Village, East Williamsburg, Financial District, Fort Greene, Fulton Ferry, Gowanus, Gramercy, Greenpoint, Greenwich Village, Hamilton Heights, Hudson Yards, Lenox Hill, Lincoln Square, Little Italy, Lower East Side, Manhattan Beach, Mill Island, Murray Hill, Noho, North Side, Paerdegat Basin, Park Slope, Prospect Heights, Prospect Park South, Red Hook, Soho, South Side, Stuyvesant Town, Tribeca, Tudor City, Upper East Side, Vinegar Hill, Washington Heights, West Village, Williamsburg, Windsor Terrace, Yorkville.

CONCLUSION - 2

• For Bronx, Queens and Staten Island

- The neighbourhoods are:
- Arlington, Arverne, Astoria, Blissville, Bloomfield, Broad Channel, Brookville, Buffer Manor, Charleston, Claremont Village, Egbertville, Elmhurst, Fieldston, Flushing, Forest Hills Garden, Fox Hills, Glendale, Grymes Hills, High Bridge, Holliswood, Howland Hook, Huguenot, Hunters Point, Hunts Point, Kew Gardens, Kingsbridge Heights, Lighthouse Hill, Midland Beach, Neoponsit, New Brighton, North Riverdale, Oakwood, Pleasant Plains, Pomonok, Port Ivory, Queensboro Hill, Randall Manor, Ravenswood, Rossville, Roxbury, Silver Lake, Somerville, South Beach, South Jamaica, South Ozone Park, Springfield Gardens, Steinway, Throgs Neck, Todt Hill, Tottenville, Travis, Utopia, West Farms.