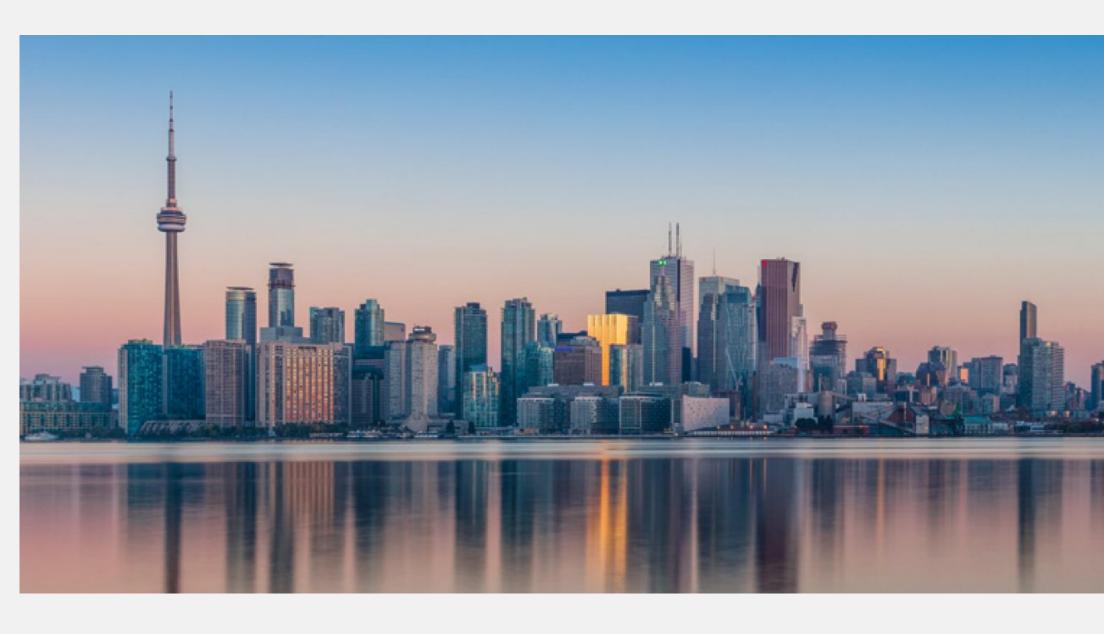
Where is a good place to open a restaurant in Toronto?





BUSINESS IDEA

WHERE TO OPEN A RESTAURANT IS ALWAYS A HARD PROBLEM FOR EVERY ENTREPREUNER, ESPECIALLY IN BIG CITY, HOW TO FIND A IDEAL PLACE WHICH NOT CLOSE TO TOO MANY COMPETITOR BUT CAN ALSO ATTRACT CUSTOMER. IN THIS ANALYSIS WE WILL COVER ON THIS PROBLEM, AND PUT A SPECIFIC TYPE AS OUR CORE TOPIC - JAPANESE RESTAURANT

Technique & Requirement

TOPIC - JAPANESE RESTAURANT

Japanese restaurant is always a competitive type of restaurant in western country, this will offer enough sample for us.

MACHINE LEARNING TECHNIQUE USED- K-MEANS

Its very straightforward and easy to achieve, it is also sensitive to noise and for this situation I think K-means is a good fit

DATA SOURCE

1.

Taipei neighborhoods data from Wikipedia

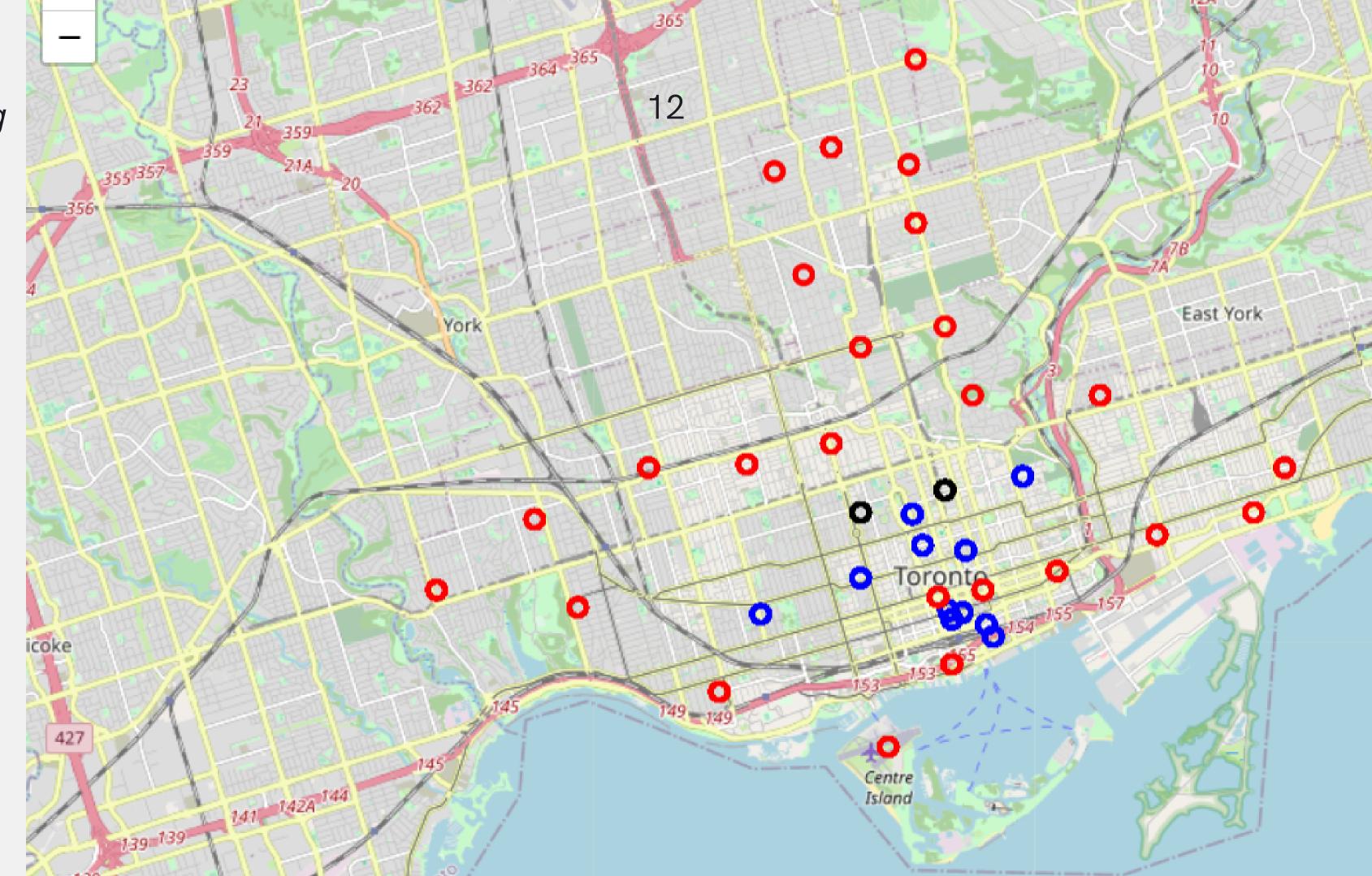
2.

Latitude and longtitude from Geocoder package

3.

Venue related data from Foursquare API

K-means
Clustering
Result



Method:

Digging into each cluster

Observation:

Clearly, we can see that most of the Japanese restaurant are in cluster 0, so we should avoid this area since it would be very risky and competitive

.6]:	#CLuster 0			
	to_merged.loc[(to_merged['Cluster	Labels'] ==0) & (to_merged['Venu	ue Category'] == 'Japanese Restaurant')]	ı

Longitude

-79.404884

-79.384707

-79.420521

-79.378891

-79.384192

-79.378788

-79.378782

-79.379608

-79.384863

-79.368415

-79.379608

-79.375887

-79.378782

-79.379608

-79.375887

-79.392494

-79.378782

-79.387424

-79.386457

43.646473

Category

Restaurant

Restaurant

Japanese

Restaurant

Restaurant

Restaurant

Japanese

Restaurant

Japanese

Restaurant

Restaurant

Restaurant

Restaurant

Restaurant

Japanese

Restaurant

Restaurant

Restaurant

Restaurant

Restaurant

Restaurant

Restaurant

ged.loc[(to[merged['Cluster Labels'] ==0) & (to_merged['Venue Category'] == 'Japanese Restaurant')							merged.ioc[(to[merged[ti	
enue Itude		Venue	Neighborhood Longitude	Neighborhood Latitude	Cluster Labels	Japanese Restaurant	Neighborhood	
2258	43.652	Gushi	-79.400049	43.653206	0	0.017241	Kensington Market, Chinatown, Grange Park	17
5085	43.668	Takyo Grill	-79.389494	43.662301	0	0.027778	Queen's Park, Ontario Provincial Government	23
8535	43.648	Bazara	-79.419750	43.647927	0	0.022727	Little Portugal, Trinity	19
0596	43.660	Kinka Izakaya Original	-79.378937	43.657162	0	0.030000	Garden District, Ryerson	13
7457	43.657	KAKA	-79.378937	43.657162	0	0.030000	Garden District, Ryerson	13
9860	43.659	Katsuya	-79.378937	43.657162	0	0.030000	Garden District, Ryerson	13
8473	43.646	Chotto Matte	-79.381576	43.647177	0	0.030000	Toronto Dominion Centre, Design Exchange	37
7167	43.647	Ki Modern Japanese + Bar	-79.381578	43.647177	0	0.030000	Toronto Dominion Centre, Design Exchange	37
0228	43.650	Ninki Izakaya	-79.381578	43.647177	0	0.030000	Toronto Dominion Centre, Design Exchange	37
5895	43.665	Kingyo Toronto	-79.387675	43.667967	0	0.023810	St. James Town, Cabbagetown	30
7167	43.647	Ki Modern Japanese + Bar	-79.374846	43.646435	0	0.032258	Stn A PO Boxes	31
0853	43.650	NAMI	-79.374846	43.646435	0	0.032258	Stn A PO Boxes	31
6473	43.646	Chotto Matte	-79.374846	43.646435	0	0.032258	Stn A PO Boxes	31
7167	43.647	Ki Modern Japanese + Bar	-79.379817	43.648198	0	0.030000	Commerce Court, Victoria Hotel	7
0853	43.650	NAMI	-79.379817	43.648198	0	0.030000	Commerce Court, Victoria Hotel	7
8008	43.656	Omai	-79.387383	43.657952	0	0.032258	Central Bay Street	4
8473	43.646	Chotto Matte	-79.379817	43.648198	0	0.030000	Commerce Court, Victoria Hotel	7
4918	43.654	Rolltation	-79.387383	43.657952	0	0.032258	Central Bay Street	4
	40.044	Fune Japanese	70.000000	40.040400	_	0.040000	First Canadian Place.	

0.040000

0.017241

Underground city

Berczy Park

43.648429

43.644771

-79.373308

Observation:

There are several Japanese in cluster1 and 2, which are both significant lower than cluster 0, so we can say if we want to open a Japanese restaurant, we should focus on cluster 1 and

#Cluster 1

to_merged.loc[(to_merged['Cluster Labels'] ==1) & (to_merged['Venue Category'] == 'Japanese Restaurant')]

	Neighborhood	Japanese Restaurant	Cluster Labels	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Ven Longitu
29	St. James Town	0.012500	1	43.651494	-79.375418	Gyu-Kaku Japanese BBQ	43.651422	-79.3750
14	Harbourfront East, Union Station, Toronto Islands	0.010000	1	43.640816	-79.381752	Miku	43.641374	-79.3775
25	Richmond, Adelaide, King	0.010989	1	43.650571	-79.384568	Fune Japanese Restaurant	43.648514	-79.3864

#Cluster 2

 $to_merged.loc[(to_merged['Cluster\ Labels']\ ==2)\ \&\ (to_merged['Venue\ Category']\ ==\ 'Japanese\ Restaurant')\]$

		Neighborhood	Japanese Restaurant	Cluster Labels	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude
k	6	Church and Wellesley	0.064935	2	43.665860	-79.383160	Kokoni Izakaya	43.664181	-79.380258
d	38	University of Toronto, Harbord	0.058824	2	43.662696	-79.400049	Yasu	43.662837	-79.403217
	38	University of Toronto, Harbord	0.058824	2	43.662696	-79.400049	Gyubee	43.667088	-79.400571
	6	Church and Wellesley	0.064935	2	43.665860	-79.383160	Tokyo Grill	43.665085	-79.384707
	6	Church and Wellesley	0.064935	2	43.665860	-79.383160	Kawa Sushi	43.663894	-79.380210
	6	Church and Wellesley	0.064935	2	43.665860	-79.383160	Onnki Donburi	43.669757	-79.384574
	6	Church and Wellesley	0.064935	2	43.665860	-79.383160	Tokyo Kitchen	43.668783	-79.385153

Cluster 3

to_merged.loc[(to_merged['Cluster Labels'] ==3) & (to_merged['Venue Category'] == 'Japanese Restaurant')]