

# Capstone Project

## Finding the best location in Toronto to open a Chinese food market

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### Agenda

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## Part 1: Intro/Business problem

Mr. Chen is a new immigrant who owned a chain supermarket company back in China.

He would like to apply his business talents to the Canadian market and open a supermarket targeting Chinese immigrants.

With everything else prepared, he now needs to know which community is the best choice for him to start his business.

His requirements for the ideal place are:

1. large Chinese population;
2. few competitors;
3. reasonable rent.

## Part 2: Data and Methodology

To meet Mr. Chen's requirements, we need both location data and demographic data for the Toronto area. Hence the 3 main data sources will be Toronto public data portal, Foursquare and CBRE.

The whole research will be carried out in 4 main steps.

First, we use Toronto data to target the main candidate regions. We collect data from Toronto public data portal to find out the most populated regions. We need pandas and matplotlib. Second, we use data from Foursquare and decide on the number of competitors and population coverage per competitor. We need folium for visualization. Third, we got data from CBRE to see the leasing costs in shortlisted regions. Fourth, we crunch relevant data into one data frame, and import reference data, which is used to do supervised machine learning in deciding which region has the best chance for success.

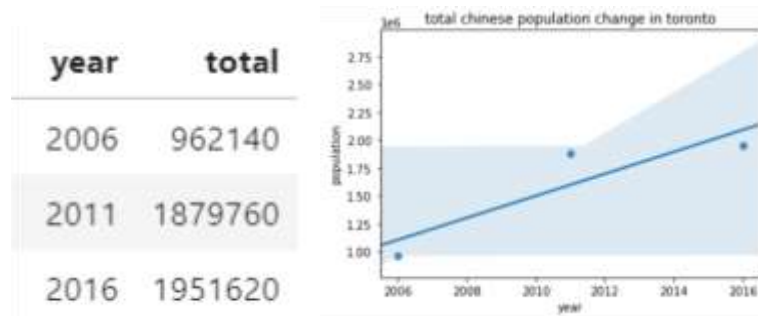
## Part 3: Finding areas most populated by Chinese immigrants

This will be solved by leveraging the demographic data from public data portal for Toronto(<https://www.toronto.ca/city-government/data-research-maps/neighbourhoods-communities/neighbourhood-profiles/>) in the most recent years. First is to locate the promising areas where lots of Chinese people reside. We will look at two numbers; one is the total number of Chinese population in each region. Second, we will need these numbers for the most recent 3 years, as a reference to the demographic trend in these areas.

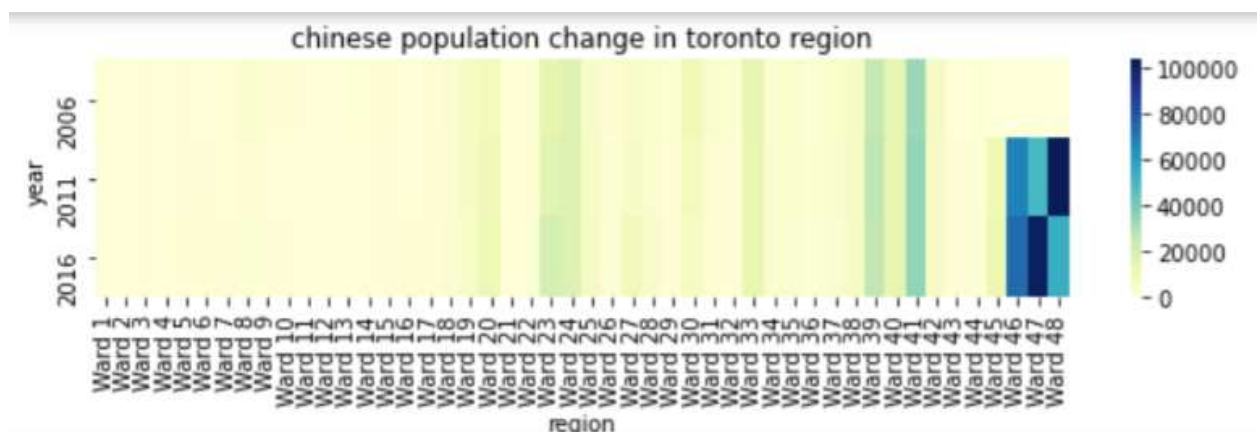
There are 48 wards in Toronto. Data for 3 years are found: 2006, 2011, and 2016. By adding up all related Chinese language we got a total population in each ward, and then putting the total population of each ward of those 3 years into one data frame. The result is as follow:

	Ward 1	Ward 2	Ward 3	Ward 4	Ward 5	Ward 6	Ward 7	Ward 8	Ward 9	Ward 10	...	Ward 39	Ward 40	Ward 41	Ward 42	Ward 43	Ward 44	Ward 45	Ward 46	Ward 47	Ward 48
year																					
2006	1125.0	665.0	1530.0	900.0	1250.0	1070.0	1145.0	3805.0	1645.0	1480.0	...	26315.0	16000.0	33980.0	6935.0	1105.0	2085.0	0.0	0.0	0.0	0.0
2011	665.0	610.0	1410.0	885.0	1360.0	1230.0	1110.0	2545.0	1470.0	1170.0	...	27305.0	15390.0	35685.0	6180.0	1260.0	1950.0	11900.0	68760.0	50450.0	103840.0
2016	800.0	563.0	1365.0	935.0	1745.0	1715.0	1345.0	2100.0	1360.0	1620.0	...	26995.0	15185.0	35505.0	5955.0	1395.0	2000.0	12760.0	74625.0	102130.0	54425.0

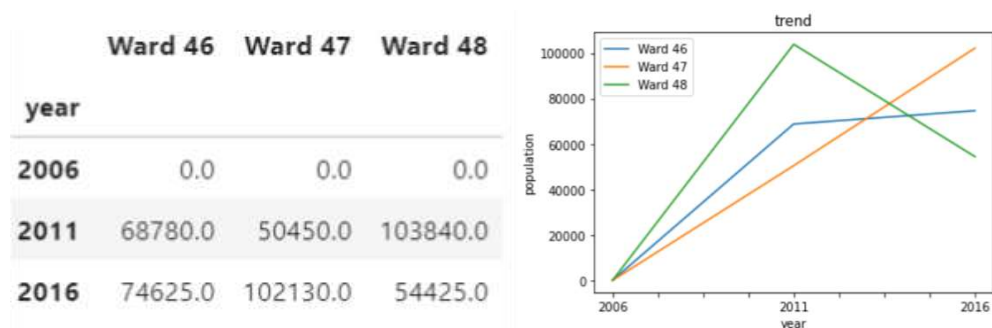
After calculation, we see that the total Chinese population in Toronto is in an increasing trend in recent years, which is a good sign for launching new business.



To make these data more readable, I plot them in a heat map, which clearly shows which region has the largest Chinese population.



Obviously the last 3 wards are more populated than other places. We will select these 3 wards as candidates. Let's explore the data further by putting the data of these wards into another plot, in order to see the trend of population change.



See in the data frame that 2006 data is missing, but it doesn't matter a lot. When referring back to the original data from Toronto portal, we know that Ward 46 is North York, Ward 47 is East York and Ward 48 is Scarborough. The plot above shows that in both North York and East York, the Chinese population is growing, and East York is having a rapid growth. While in Scarborough the Chinese population is decreasing sharply. Nevertheless, Scarborough still has a large Chinese population, so it is still shortlisted.

## Part 4: Decide on area that is least competitive

Foursquare will be the main source of data for this step. In the previous step we get a list of wards with large and growing Chinese population. Using the geographical info of those wards, we put it into Foursquare and explore those areas.

We use geolocator to decide the ward centers' latitude and longitude. According to the acreages of the shortlisted wards, the searching radius of North York and Scarborough is set to be 5 km while that for East York 2 km.

All Asian markets are considered as competitors. So when doing research on Foursquare, the searching keyword is set to be "Asian Market". Below are the 3 dataframes and corresponding maps showing the competitors' information in each ward.

North York:

	name	categories	lat	lng	labelLatLng	distance	is	country	formattedAddress	address	postalCode	city	state	crossStreet	neighborhood	id
1	Whole Foods Market	Grocery Store	43.76952	-79.411044	[[{"lat": 43.76952, "lng": -79.411044}]]	3141	CA	Canada	[4771 Yonge Street, North York, ON M2N 5M5, Canada]	4771 Yonge Street	M2N 5M5	North York	ON	NaH	NaH	579b7b2a87d1ac5e8b34a99
2	Asian Gourmet	Asian Restaurant	43.790940	-79.422880	[[{"lat": 43.79094, "lng": -79.42288}]]	3089	CA	Canada	[Centropoint, Toronto, ON, Canada]	Centropoint	NaH	Toronto	ON	NaH	NaH	4a4743e675d72af7a6b9c5c4e
7	Asian Cuisine	Asian Restaurant	43.811545	-79.454070	[[{"lat": 43.811545, "lng": -79.45407}]]	6399	CA	Canada	[Vaughan, ON, Canada]	NaH	NaH	Vaughan	ON	NaH	NaH	4c4a5f181797236a5a9d5086
11	Coppel's Fresh Market	Supermarket	43.776956	-79.489547	[[{"lat": 43.776956, "lng": -79.489547}]]	3057	CA	Canada	[4750 Dufferin St (at Martin Rd), North York, ON M3H 5T7, Canada]	4750 Dufferin St	M3H 5T7	North York	ON	at Martin Rd	NaH	4b4a5a18964a537a85a25a5
18	Seniors Fresh Market	Market	43.718137	-79.429905	[[{"lat": 43.718137, "lng": -79.429905}]]	4344	CA	Canada	[Canada]	NaH	NaH	NaH	NaH	NaH	NaH	579a882496a525815a0724f
19	DPV Marie's Food Market	Grocery Store	43.737232	-79.436557	[[{"lat": 43.737232, "lng": -79.436557}]]	2761	CA	Canada	[322 Wilson Ave (Collinson Blvd), North York, ON M2N 5M5, Canada]	322 Wilson Ave	M2N 5M5	North York	ON	Collinson Blvd	NaH	5a5087e02a2a5a000b8a9e5
20	Bathurst Village Market	Supermarket	43.784963	-79.445884	[[{"lat": 43.784963, "lng": -79.445884}]]	3319	CA	Canada	[Toronto, ON M2N 1Z1, Canada]	NaH	M2N 1Z1	Toronto	ON	NaH	NaH	5c395a4a5439002c5a4a55
23	Heal Research (market research)	Home	43.767594	-79.411041	[[{"lat": 43.767594, "lng": -79.411041}]]	3326	CA	Canada	[5075 Yonge St, Toronto, ON, Canada]	5075 Yonge St	NaH	Toronto	ON	NaH	NaH	4a800e1124a5a8a5a5a5a5



Scarborough (dataframe is abridged):

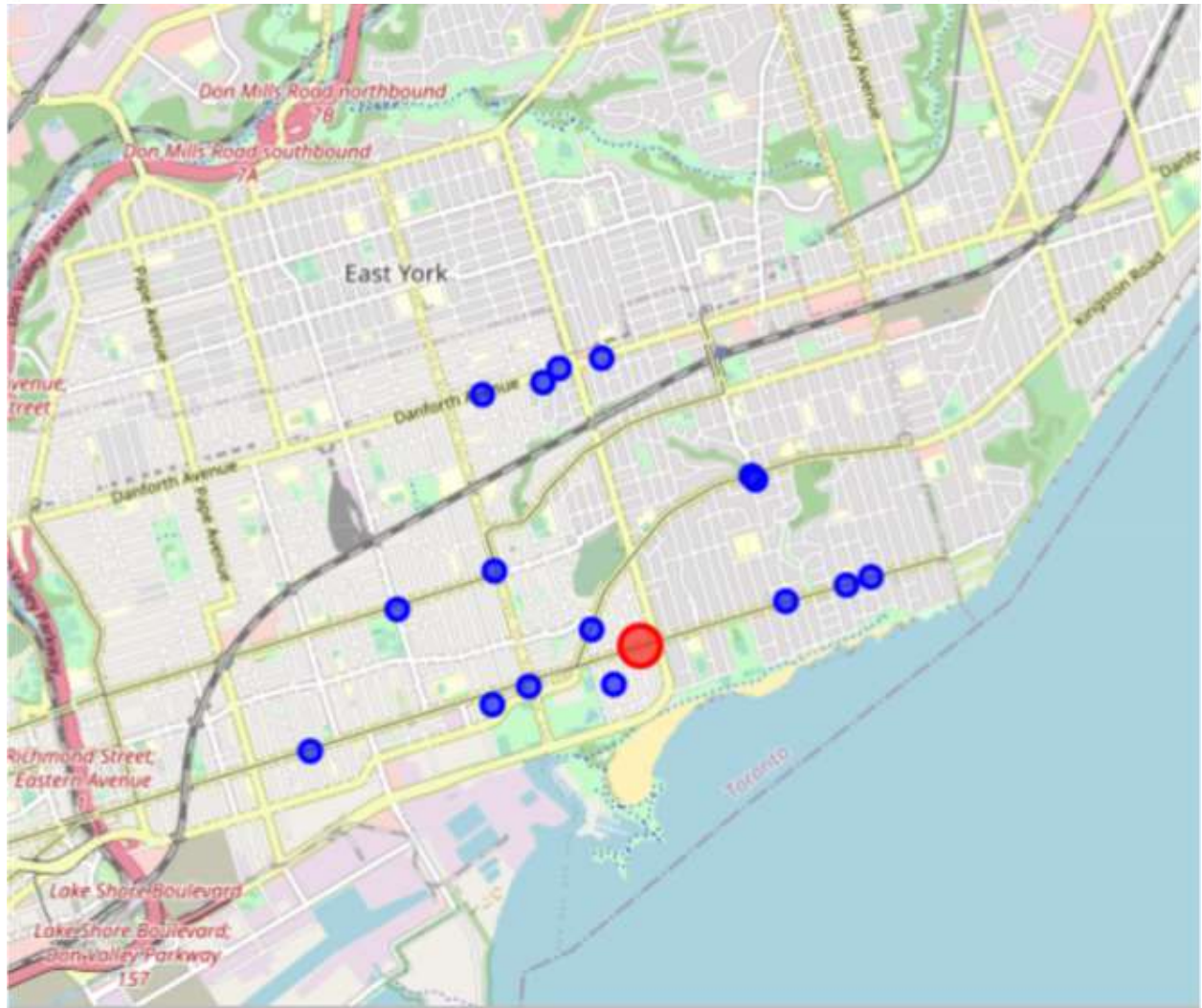
	name	categories	address	crossStreet	lat	lng	labeledLatLngs	distance	cc	city	state	country	formattedAddress	postalCode	neighborhood	id
4	Asian Specialty Restaurant	Vietnamese Restaurant	2301 Brimley Rd.	at Huntingwood Dr.	43.797819	-79.270737	[[{"label": "display", "lat": 43.797819, "lng": ...}], [{"label": "crossStreet", "lat": 43.797819, "lng": ...}]]	2044	CA	Toronto	ON	Canada	[[2301 Brimley Rd. at Huntingwood Dr.], Toronto...	M1S 5B8	Na/N	4f79fcd64b0c96c9ef99f5d
5	Asian Gourmet	Chinese Restaurant	Lawrence	Markham	43.756263	-79.127862	[[{"label": "display", "lat": 43.756263, "lng": ...}], [{"label": "crossStreet", "lat": 43.756263, "lng": ...}]]	2915	CA	Toronto	ON	Canada	[[Lawrence (Markham), Toronto ON, Canada]	Na/N	Na/N	4c148e382a5c9b60009he9
7	Jody's Asian Cuisine	None	Na/N	Na/N	43.795086	-79.260483	[[{"label": "display", "lat": 43.795086, "lng": ...}], [{"label": "crossStreet", "lat": 43.795086, "lng": ...}]]	1354	CA	Na/N	Na/N	Canada	[[Canada]	Na/N	Na/N	49021289a60454619ab5ea7
12	Edgcomb's Marketing And Promotions (warehouse)	Office	2250 Midland Ave. Toronto, ON M1P 3E5	Na/N	43.775695	-79.274705	[[{"label": "display", "lat": 43.775695, "lng": ...}], [{"label": "crossStreet", "lat": 43.775695, "lng": ...}]]	1391	CA	Toronto	ON	Canada	[[2250 Midland Ave. Toronto, ON M1P 3E5, Toronto...	M1P 3E5	Na/N	48f19361a6b22a9583ed0ba7
20	Kostas Meat Market	Greek Restaurant	229 Elsomere Rd	Warden Ave	43.760605	-79.301850	[[{"label": "display", "lat": 43.760605, "lng": ...}], [{"label": "crossStreet", "lat": 43.760605, "lng": ...}]]	3854	CA	Toronto	ON	Canada	[[229 Elsomere Rd (Warden Ave), Toronto ON M1R...	M1R 4B4	Na/N	4a0f8b6c194845cfa2b1aef6f
21	Adlytic Marketing	Advertising Agency	2190 Elsomere Road	Na/N	43.777470	-79.230301	[[{"label": "display", "lat": 43.7774699919856...}], [{"label": "crossStreet", "lat": 43.7774699919856...}]]	2269	CA	Toronto	ON	Canada	[[2190 Elsomere Road Toronto ON M1G 3M5, Canada]	M1G 3M5	Scarborough	1e4a0b67e1617b00035ca334
23	Marketplace Ministries	Church	1776 Midland Ave	Midland And Elsomere	43.762053	-79.270001	[[{"label": "display", "lat": 43.762053, "lng": ...}], [{"label": "crossStreet", "lat": 43.762053, "lng": ...}]]	1576	CA	Scarborough	ON	Canada	[[1776 Midland Ave (Midland And Elsomere), Scar...	M1P 3P2	Na/N	40f1c3c2e4b0aef05556c3a0f





East York (dataframe is abridged):

	name	categories	address	crossStreet	lat	lng	labeledLatLngs	distance	postalCode	cc	city	state	country	formattedAddress	id
1	raise the root organic market	Fruit & Vegetable Store	1154 Queen Street east	Dover Avenue	43.661180	-79.309204	[[{"lat": 43.661180, "lng": 43.661180, "display": "lat"}, {"lat": 43.661180, "lng": 43.661180, "display": "lng"}]]	213	M4V 1A1	CA	Toronto	ON	Canada	[1154 Queen Street east (Dover Avenue), Toronto, ON M4V 1A1, Canada]	54b1a70a438a4d4710c4b0
2	Beaches Wholesale Market	Health Food Store	2234 Queen St E	Beach Ave	43.672581	-79.287557	[[{"lat": 43.672581, "lng": 43.672581, "display": "lat"}, {"lat": 43.672581, "lng": 43.672581, "display": "lng"}]]	1548	M4E 1G2	CA	Toronto	ON	Canada	[2234 Queen St E (Beach Ave), Toronto, ON M4E 1G2, Canada]	4c547d8995d13af5a1c3afa
3	Today's Food Market	Convenience Store	86 Kingston Rd		43.669536	-79.211346	[[{"lat": 43.669536, "lng": 43.669536, "display": "lat"}, {"lat": 43.669536, "lng": 43.669536, "display": "lng"}]]	348	M4E 1G6	CA	Toronto	ON	Canada	[86 Kingston Rd, Toronto, ON M4E 1G6, Canada]	4da1ec12800e117a27ef258
4	Leslieville Farmers Market	Farmers Market			43.664901	-79.319794	[[{"lat": 43.664901, "lng": 43.664901, "display": "lat"}, {"lat": 43.664901, "lng": 43.664901, "display": "lng"}]]	1088		CA	Toronto	ON	Canada	[Toronto, ON, Canada]	4db0ba688774880a33870d
5	The Big Carrot Natural Food Market	Health Food Store	125 Southwood Dr		43.678879	-79.297754	[[{"lat": 43.678879, "lng": 43.678879, "display": "lat"}, {"lat": 43.678879, "lng": 43.678879, "display": "lng"}]]	1280	M4E 0B8	CA	Toronto	ON	Canada	[125 Southwood Dr, Toronto, ON M4E 0B8, Canada]	4a44062964e1201182be3
7	Farmer's Fruit Market	Farmers Market			43.670201	-79.218814	[[{"lat": 43.670201, "lng": 43.670201, "display": "lat"}, {"lat": 43.670201, "lng": 43.670201, "display": "lng"}]]	1107		CA	Toronto	ON	Canada	[Toronto, ON, Canada]	4b0906e48c008ba7400e11
8	The Dog Market	Pet Store	2116 Queen St E	HammerSmith Ave	43.671163	-79.294773	[[{"lat": 43.671163, "lng": 43.671163, "display": "lat"}, {"lat": 43.671163, "lng": 43.671163, "display": "lng"}]]	1044	M4E 1E2	CA	Toronto	ON	Canada	[2116 Queen St E (HammerSmith Ave), Toronto, ON M4E 1E2, Canada]	4a44062964e1201182be3
11	The Big Carrot Natural Food Market	Grocery Store	125 Southwood Dr		43.678879	-79.297754	[[{"lat": 43.678879, "lng": 43.678879, "display": "lat"}, {"lat": 43.678879, "lng": 43.678879, "display": "lng"}]]	1278	M4E 0B8	CA	Toronto	ON	Canada	[125 Southwood Dr, Toronto, ON M4E 0B8, Canada]	4a44062964e1201182be3
12	Rowan Homegrown Market	Market	2196 Queen St E	Balmain Ave	43.672218	-79.289780	[[{"lat": 43.672218, "lng": 43.672218, "display": "lat"}, {"lat": 43.672218, "lng": 43.672218, "display": "lng"}]]	1463	M4E 1E6	CA	Toronto	ON	Canada	[2196 Queen St E (Balmain Ave), Toronto, ON M4E 1E6, Canada]	4a44062964e1201182be3



By using dataframe.shape we can conclude that North York has 8 competitors, Scarborough has 8, and East York has 16. The number of competitors is not representing the full picture, because different ward has different population. The number will be meaningful only if we factor in the population. That is, divide the population by competitor number to see how large is the population each competitor cover, which we will name it as population coverage in the following part. The larger is the population coverage, the larger potential is the market, because that means each potential customer gets less choice.

Now let's see what happens if we do the population coverage calculation for the 3 shortlisted wards:

	ward	population	competitors number	population coverage per competitor
0	North York	74625	8	9328.125
1	Scarborough	54425	8	6803.125
2	East York	102130	16	6383.125

The above dataframe lists 3 wards in descending order for their population coverage, with North York having the biggest population coverage.

## Part 5: Decide on area with lightest leasing cost

Apart from the population coverage, Mr. Chen also wants to know the leasing cost, and uses both factors to determine the location choice.

CBRE will be the data source for getting this information. CBRE offers clear data report about business rental information for the Toronto area. Here is a snapshot showing the leasing costs of the 3 wards:

Submarket	Inventory (SQ)	Vacancy Rate (%)	Sublet Space (% of Vacant Space)	Net Absorption (SQ)	YTD Net Absorption (SQ)	New Supply (SQ)	Under Construction (SQ)	Leasing Rate (\$/SQ)
129700	87,781,242	1.9	18.7	1,087,117	1,087,117	0	0,000,000	33.85
129700	75,223,000	2.0	17.0	750,000	750,000	0	0,000,000	33.85
Financial Core	34,432,000	3.6	15.0	33,300	33,300	0	0,000,000	40.00
Midtown Core	19,434,114	1.8	30.0	19,300	19,300	0	3,000,000	30.76
Greenwich South	4,415,100	0.9	30.1	12,100	12,100	0	0,000,000	31.18
Greenwich North	7,000,000	0.1	30.0	20,000	20,000	0	0,000,000	31.90
Greenwich East	3,137,000	1.6	18.0	19,000	19,000	0	0,000,000	39.47
Greenwich West	3,600,750	3.4	18.1	14,000	14,000	0	0,000,000	31.80
Liberty Village	3,240,000	0.1	10.0	1,100	1,100	0	0,000,000	37.10
Midtown	11,011,111	1.0	11.0	100,000	100,000	0	0	30.76
West / Finch	7,300,220	1.1	7.0	79,000	79,000	0	0	30.80
St. Clair / Finch	2,190,000	1.0	0.0	0,000	0,000	0	0	34.70
Highway 7 / Finch	4,000,000	2.0	10.1	40,000	40,000	0	0	30.80
129700	11,100,000	1.0	10.0	100,000	100,000	0,000,000	0,000,000	34.80
129700	10,000,000	1.0	10.0	100,000	100,000	0	0	34.80
Scarborough	3,300,000	14.0	0.0	0,000	0,000	0	0	33.70
Midtown N. / E. Hill	1,575,000	1.0	10.0	150,000	150,000	0	0	34.40
Midtown South	4,000,000	1.0	0.0	0,000	0,000	0	0	31.80
E. York / St. Andrew	1,000,000	0.0	0.0	0,000	0,000	0	0	33.70
Don Mills North	2,000,000	1.0	0.0	0,000	0,000	0	0	34.70
Commerce Court	1,000,000	0.0	0.0	0,000	0,000	0	0	34.70
St. Clair / W. York	1,000,000	0.0	0.0	0,000	0,000	0	0	34.70
North	1,000,000	1.0	0.0	0,000	0,000	0	0	31.70
North York	1,000,000	1.0	0.0	0,000	0,000	0	0	31.70
North York West	1,000,000	1.0	0.0	0,000	0,000	0	0	31.70
North York	1,000,000	1.0	0.0	0,000	0,000	0	0	31.70

(<https://www.cbre.ca/en/research-and-reports/Toronto-Office-MarketView-Q1-2020>)

The corresponding leasing price for the 3 wards are included in the following chart with the population coverage information:

	ward	population coverage per competitor	leasing cost
0	North York	9328.125	21.56
1	Scarborough	6803.125	12.20
2	East York	6383.125	12.52

## Part 6: Modelling and Testing

Now we have the information that Mr. Chen requires at hand. We can start our analysis.

The final answer we want to get is whether the business will be successful or not, by analyzing the information we got. This is very similar to a prediction model in machine learning. Hence classification will be the method we use. We will try to use data features to predict the business in each ward will be successful (labelled as Y) or not (labelled as N).

We found some historical data for supermarkets running in Toronto areas, which can be used to train the machine learning model. We use train-split-test to get the best model and then apply the model to real data for prediction.

Here is the data for modelling:

	Name	population coverage per competitor	leasing cost	Success or not
0	Market 1	10500	27	N
1	Market 2	7500	20	N
2	Market 3	6800	10	Y
3	Market 4	6600	9	Y
4	Market 5	7700	15	Y
5	Market 6	8000	13	Y
6	Market 7	9900	28	N
7	Market 8	11000	24	Y
8	Market 9	9800	19	Y
9	Market 10	8800	23	N

For the convenience of modelling, we change the Y/N values into 1/0 as follow:



	Name	population	coverage per competitor	leasing cost	Success or not
0	Market 1		10500	27	0
1	Market 2		7500	20	0
2	Market 3		6800	10	1
3	Market 4		6600	9	1
4	Market 5		7700	15	1
5	Market 6		8000	13	1
6	Market 7		9900	28	0
7	Market 8		11000	24	1
8	Market 9		9800	19	1
9	Market 10		8800	23	0

We use 80% of the data to train the model and 20% to test it. We process the “competitor” and “leasing cost” into numpy array as X, “success or not” as Y. KNN will be the method for use. We then use accuracy metrics to test how accurate it is. Setting K=2, we got the following result:

```
[51]: from sklearn.neighbors import KNeighborsClassifier
      #Start testing by using k as 2.
      k = 2
      #Train Model and Predict
      neigh = KNeighborsClassifier(n_neighbors = k).fit(X_train,Y_train)
      neigh

/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages/ipykernel_launcher
the shape of y to (n_samples, ), for example using ravel().
"""
[51]: KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
      metric_params=None, n_jobs=None, n_neighbors=2, p=2,
      weights='uniform')

[52]: #We use the model to predict the test result.
      Yhat = neigh.predict(X_test)
      Yhat

[52]: array([1, 0])
```

Yhat is 1 and 0, by referring back to original data, the prediction is correct. Accuracy for train set is 0.875 and test set 0.5.

The result looks nice, but we want to find the best model, so we try to test every K value possible, from 1 to 9. The calculation tells us that the best K is 1, when train set accuracy is 1 and test set 0.5. Then we redo the model with k=1 and apply the model to actual data of the 3 shortlisted wards.

```
[57]: #so the best k is 1. Redo the model.
      k = 1
      neigh = KNeighborsClassifier(n_neighbors = k).fit(X_train,Y_train)
      neigh

/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages/ipykernel_launcher.py:3: DeprecationWarning:
the shape of y to (n_samples, ), for example using ravel().
This is separate from the ipykernel package so we can avoid doing imports until

[57]: KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
                          metric_params=None, n_jobs=None, n_neighbors=1, p=2,
                          weights='uniform')

[58]: #imput the data that we need to predict for the 3 wards.
      X1=df[['population coverage per competitor','leasing cost']].values
      X1=preprocessing.StandardScaler().fit(X1).transform(X1.astype(float))

[59]: #now Let's see using the KNN model we have how's the result for each ward.
      Y1hat = neigh.predict(X1)
      Y1hat

[59]: array([0, 1, 1])
```

It gives a final result as an array [0,1,1] for the 3 wards.

## Part 7: Result

We transform that Boolean values back into Y/N values and integrate them into the data frame:

	ward	population coverage per competitor	leasing cost	Success or not
0	North York	9328.125	21.56	N
1	Scarborough	6803.125	12.20	Y
2	East York	6383.125	12.52	Y

The analysis for the 3 wards has a preliminary conclusion:

To open a Chinese market in the 3 wards, North York will be not successful, Scarborough and East York will be successful.

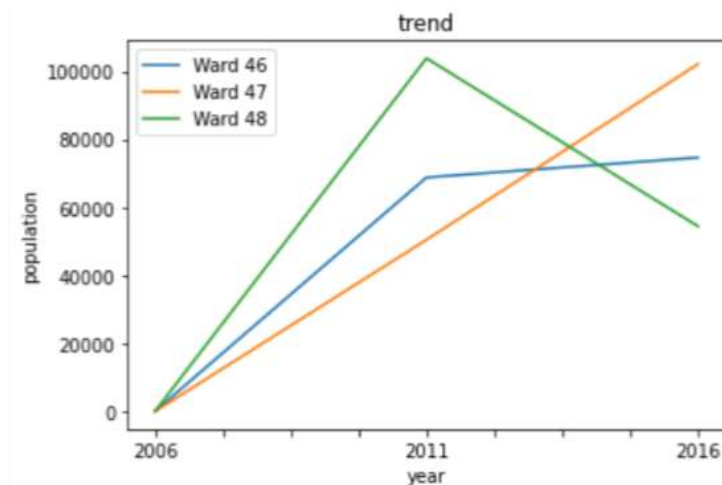
## Part 8: Discussion

Unfortunately, North York fails the test. The leasing cost is too high that the population coverage cannot justify it. For the other 2 wards, does it mean that both Scarborough and East York have the same advantages for opening a Chinese market?

Of course there is difference. First we take a look at the straightforward information showing by charts and plots.

In the result chart in Part 7, we can see that in population coverage, Scarborough slightly wins over East York (6803 over 6383, the difference is about 5%). For the leasing costs, too, Scarborough wins by 0.32 CAD per square foot (12.20 compared with 12.52 in East York, the difference is about 2%).

This is the current situation. However, opening a business needs to look into the future. As per the result we got from the population analysis part, we can see that the total population in Toronto is in an increasing trend, hence we suppose the population coverage and leasing cost will increase with it. However, something special about Toronto is that it is a diverse city with people of different ethnic origins tend to live in the same region, so we have China town, little Portugal, little Italy etc. There is a trend for the region population change, i.e., if there are people starting leaving a certain region to live in another region, the rest population may follow the trend. Now recall the following plot:



Just to remind that Ward 47 is East York and Ward 48 is Scarborough. We can see that the Chinese population in Scarborough is decreasing sharply by 47.5%! While that in East York is growing by 102%! We can conclude that Chinese immigrants are leaving Scarborough and more like to settle down in East York.

Population is the most important feature we need to consider for potential market. Besides, when comparing with the population trend change, the difference of leasing costs and population coverage look small. Therefore, even though East York is slightly less advantageous to Scarborough in leasing cost and population coverage, the former's population trend benefit is much better than the latter. East York is a better choice.

## Part 8: Conclusion

Weighing in the population trend, population coverage and leasing cost, the final winner is:

**East York!!!**