

# Vancouver Grocery

## 1. Introduction

### 1.1. Background

From grocery shops customers buy everyday necessities for them and their families.

There's a wide range of products which for each there is a wide range of qualities and as a consequence price.

The task of finding a perfect location for a new retail shop is a very demanding one, let alone for a grocery shop.

There are several methods of finding the perfect spot, starting from the more traditional such as wandering through the roads of some neighborhoods which have been picked up either by chance or based upon some advice of some expert

More advanced method is gathering and studying publicly available data about the neighborhoods in the city of interest. Population, neighborhood demographics, visibility, the amount of traffic that goes by and local competition are all factors taken into consideration. This data will be used to study each of the following 3 aspects about what makes a location good for a retail shop (2).

### 1.2. Problem Description

The board members of our company, FreshGems, have assigned our team to research the best location for our new high-end grocery shop in Vancouver city, Canada. This new shop would help us target customers looking for high quality groceries, harvested according to the latest bio standards.

This is a new hype with many potential customers as people are increasingly aware of the benefits of proper healthy nutrition with bio ingredients. Based on our company's ethics, healthy nutrition should be within the reach of all levels of society, thus the result of our work should be a location where there is a fairly high concentration of retail shops that target the lower half of the market.

## 2. Solution approach and data mining

### 2.1. Data Requirements

From the 4 aspects that we'll study we derive the following sets of data that will be used during the study:

- Customers: popularity of grocery shops in contrast to supermarkets
- Find out which neighborhoods are high exposure ones
- Popularity of each grocery shop per neighborhood
- Combining all of the above to make a suggestion

#### 2.1.1. Customers

We'll gather data using Foursquare API about the popularity of the grocery shops and the supermarkets in each of the neighborhoods. Based upon the coordinates of each neighborhood, a "SEARCH" request is sent to the foursquare API with radius set to 500 m and a list of categories of shops that deal with food and beverages

### 2.1.2. High exposure streets / neighborhoods

We'll use the K-Means library to cluster the popularity of the neighborhoods based on the popularity of the existing shops

### 2.1.3. Popularity of groceries per neighborhood

To be able to determine centers of competition we'll study the concentration and popularity of grocery shops within 500m radius. From the data gathered in part 1 and for each of the acquired venues, a "VENUE" request is sent to the foursquare API from which the venues' details are returned along with its rating

## 3. Methodology

### 3.1. Assumptions

Due to lack of resources the data of the city of Toronto is used to conduct the study, the "SEARCH" requests of each neighborhood are merged into one dataframe which is then exported to a csv file to conserve foursquare API request quota.

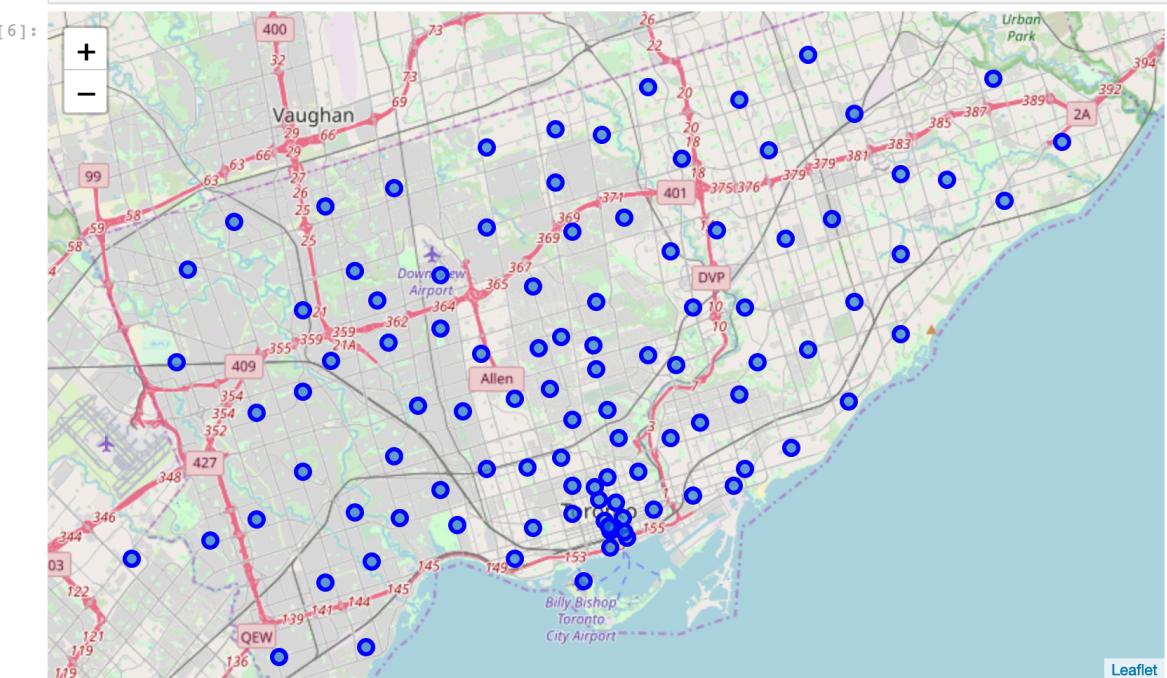
Due to the big number of venues recorded and the fact that foursquare quota is not enough to get each venue's rating, a random 2 decimal number between 0 and 10 is assigned to each.

### 3.2. Steps of the analysis

First the required libraries are imported that will be used throughout our data processing session.

Next the list of Toronto's postal codes is scrapped from Wikipedia and loaded into a dataframe.

Next the coordinates of Toronto city are acquired, and the neighborhoods are plotted on the map to visualize the area being studied.



At this point venues' data is requested from the foursquare APIs for each neighborhood and for each venue a rating is assigned.

[15]:	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category	rating
0	Rouge, Malvern	43.806686	-79.194353	Meena's Fine Foods	43.804476	-79.199753	Indian Restaurant	3.58
1	Rouge, Malvern	43.806686	-79.194353	Wendy's	43.807448	-79.199056	Fast Food Restaurant	4.86
2	Rouge, Malvern	43.806686	-79.194353	Second Cup	43.802165	-79.196114	Coffee Shop	8.01
3	Highland Creek, Rouge Hill, Port Union	43.784535	-79.160497	Shamrock Burgers	43.783823	-79.168406	Burger Joint	5.15
4	Highland Creek, Rouge Hill, Port Union	43.784535	-79.160497	Tony Stacey Centre for Veterans Care	43.783857	-79.164738	Assisted Living	8.48

One hot encoding is used to remove the numerical bias that's added by the count of similar venues in each neighborhood.

Then the data is grouped by neighborhood summing over the number of similar venue categories, aggregating the count of venues in each category.

[19]:	Neighborhood	Afghan Restaurant	African Restaurant	American Restaurant	Asian Restaurant	Assisted Living	BBQ Joint	Bagel Shop	Bakery	Bar	...	Tea Room	Thai Restaurant
0	Adelaide, King, Richmond	0	0	4	0	0	0	0	0	2	...	0	1
1	Agincourt	0	0	0	0	0	0	0	0	0	...	0	0
2	Agincourt North, L'Amoreaux East, Milliken, St...	0	0	0	2	0	3	0	2	0	...	0	0
3	Albion Gardens, Beaumont Heights, Humbergate, ...	1	0	0	0	0	0	1	5	0	...	0	0
4	Alderwood, Long Branch	0	0	0	2	0	2	0	0	0	...	0	1

5 rows x 152 columns

As a control measure the top 5 venues of each neighborhood are displayed.

```
-----Adelaide, King, Richmond-----
   venue freq
0    Coffee Shop 13.0
1 American Restaurant 4.0
2      Restaurant 3.0
3        Café 3.0
4 Ramen Restaurant 3.0
```

```
-----Agincourt-----
   venue freq
0 Sandwich Place 2.0
1 Coffee Shop 2.0
2 Chinese Restaurant 1.0
3 Korean Restaurant 1.0
4 Breakfast Spot 1.0
```

```
-----Agincourt North, L'Amoreaux East, Milliken, Steeles East-----
   venue freq
0 Chinese Restaurant 4.0
1 BBQ Joint 3.0
2 Asian Restaurant 2.0
3 Bakery 2.0
4 Fast Food Restaurant 2.0
```

The mean value of the ratings per venue type for each neighborhood is calculated.

[22]:	Neighborhood	Venue Category	rating
0	Adelaide, King, Richmond	American Restaurant	3.707500
1	Adelaide, King, Richmond	Bar	4.305000
2	Adelaide, King, Richmond	Brazilian Restaurant	3.210000
3	Adelaide, King, Richmond	Breakfast Spot	2.390000
4	Adelaide, King, Richmond	Burger Joint	5.020000
5	Adelaide, King, Richmond	Café	5.160000

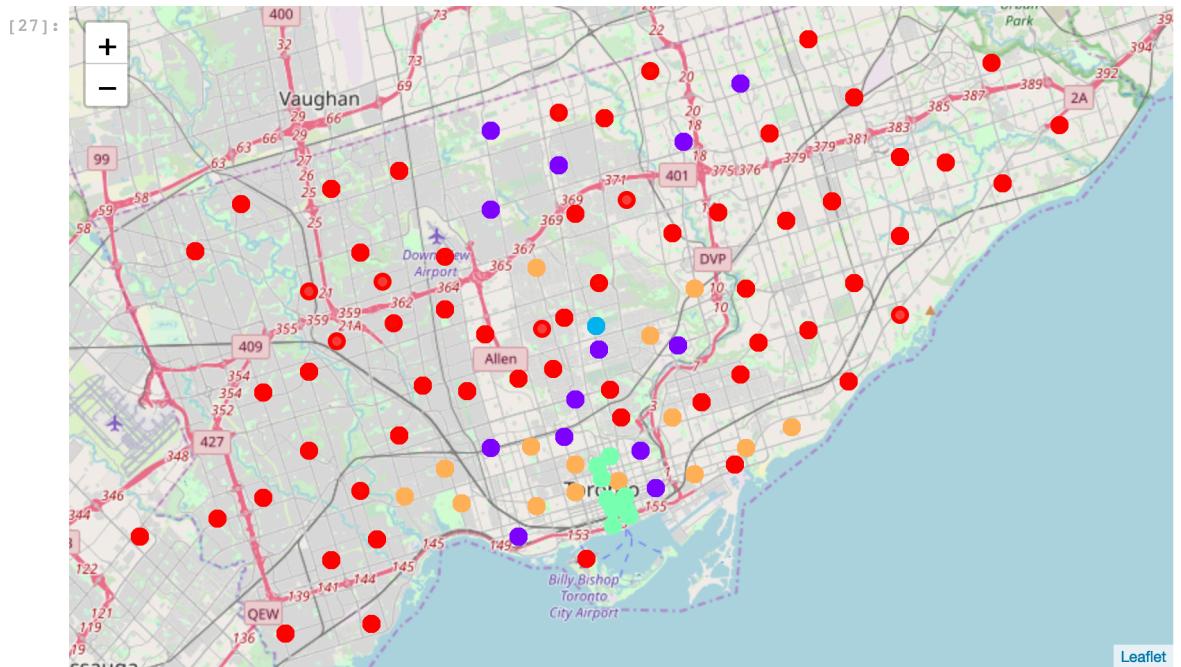
With the aid of a helper function a dataframe with the top 10 venues of each neighborhood.

[24]:	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Common Ven
0	Adelaide, King, Richmond	Coffee Shop	American Restaurant	Café	Ramen Restaurant	Restaurant	Fast Food Restaurant	Gastropub	Hotel	Bar	Br
1	Agincourt	Coffee Shop	Sandwich Place	Restaurant	Korean Restaurant	Breakfast Spot	Chinese Restaurant	Wings Joint	Dumpling Restaurant	Donut Shop	I Resta
2	Agincourt North, L'Amoreaux East, Milliken, St...	Chinese Restaurant	BBQ Joint	Fast Food Restaurant	Asian Restaurant	Bakery	Pizza Place	Medical Center	Dumpling Restaurant	Caribbean Restaurant	Ch Bre
3	Albion Gardens, Beaumont Heights, Humbergate, ...	Pizza Place	Bakery	Caribbean Restaurant	Indian Restaurant	Fast Food Restaurant	Mexican Restaurant	Residential Building (Apartment / Condo)	Chinese Restaurant	Coffee Shop	A Resta
4	Alderwood, Long Branch	Coffee Shop	Breakfast Spot	Asian Restaurant	BBQ Joint	Pizza Place	Café	New American Restaurant	Market	Deli / Bodega	Ch Resta

K-means clustering is run on the dataset that would cluster the venues into 5 groups based on each group's rating.

Venue Category	rating	Cluster Labels	1st Most Common Venue	2nd M Comm Ven
Indian Restaurant	3.58	0	Coffee Shop	Fast F
Fast Food Restaurant	4.86	0	Coffee Shop	Fast F
Coffee Shop	8.01	0	Coffee Shop	Fast F
Burger Joint	5.15	0	African Restaurant	Assis
Assisted Living	8.48	0	African Restaurant	Assis

On a map the clusters are plotted each in a separate color to visualize them.



### 3.3. Observations

The data at hand had an excellent structure and is very comprehensive, the list of postal codes is complete, the list of coordinates was used from a reputable source and was of superb quality. The foursquare API provides high quality structured data especially when the query is parameterized properly to return venues of categories only specific to the type to which this study is related. All this allowed the immediate processing of the data without further modifications.

## 4. Conclusion

From the cluster visualization it can be concluded that the center of Toronto, where there is a high number of passersby, higher population density and large number of eateries, is the ideal place to open a grocery store because firstly it's a well exposed location which would help in increasing the number of new customers daily secondly high population density implies a big potential customer base and as it's explained above, the grocery store would also benefit from surrounding grocery stores which are already a destination to get the required attention to its unique products.