

# Business Location Explore in Toronto

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# Business Location Explore in Toronto

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# 1. Introduction

- 1.1 Business Requests
- 1.2 Analytic Approach



# 1.1 Business Requests

- Generate The location problem
- Location to start a new business

We assume:

An investor wants to start a new business to serve Chinese food in downtown Toronto due to its density of population, higher average income, as well as the diversity of culture.



## 1.2 Analytic Approach

A good location should satisfy the two criteria at the same time:

- (1) Sufficient demand
- (2) Insufficient support



## 2. Data Collection

- 2.1 Scrape location info in Toronto
- 2.2 Fetch all 'FOOD' venues in Toronto
- 2.3 Show venues on map



## 2.1 Scrape location info in Toronto

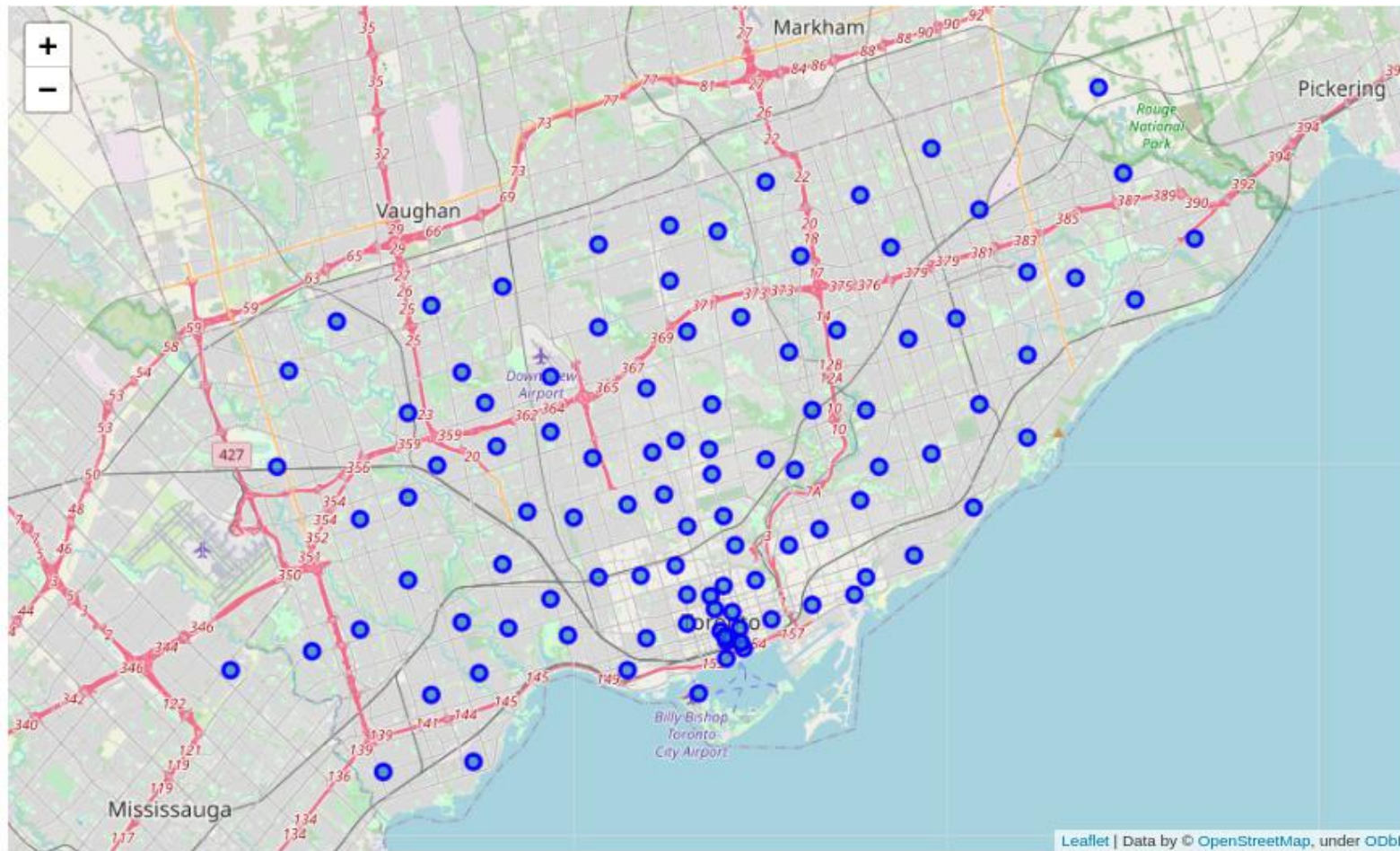
- Step(1) Fetch postal code in Toronto
- Step(2) Attaching geo info for each postal code

	Borough	Neighborhood	Latitude	Longitude
Postal Code				
M3A	North York	Parkwoods	43.753259	-79.329656
M4A	North York	Victoria Village	43.725882	-79.315572
M5A	Downtown Toronto	Regent Park / Harbourfront	43.654260	-79.360636
M6A	North York	Lawrence Manor / Lawrence Heights	43.718518	-79.464763
M7A	Downtown Toronto	Queen's Park / Ontario Provincial Government	43.662301	-79.389494



## 2.1 Scrape location info in Toronto

- Step (3) Visualization areas on map





## 2.2 Fetch all 'FOOD' venues in Toronto

- Step(1) First we set API credentials
- Step(2) Fetch data via FourSquare API
- Step(3) Review venues contains Restaurant in their Name

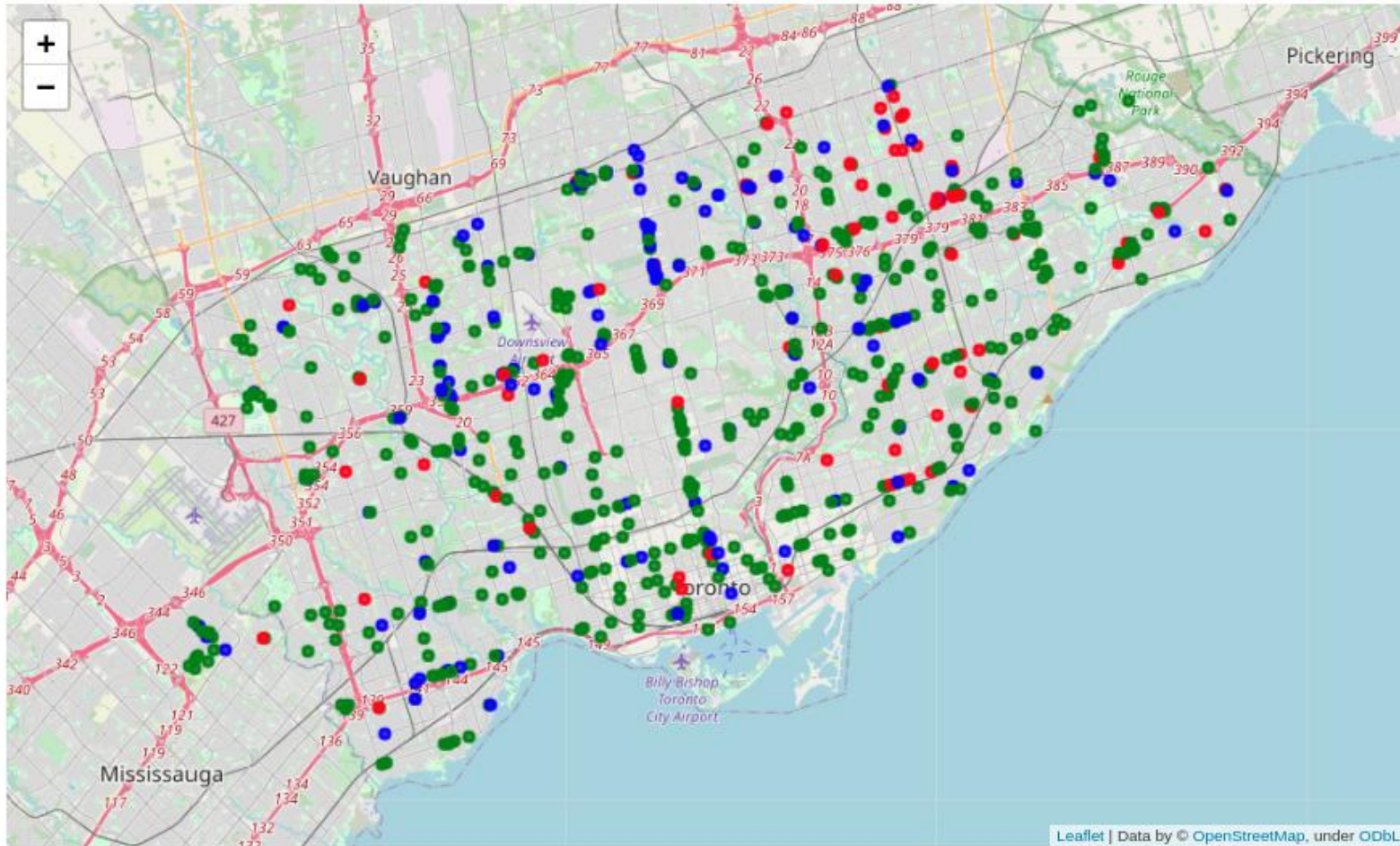


## 2.3 Show venues on map

- Step (1) Find the center of the map
- Step (2) Mark the venues on maps



# Mark the Restaurants on maps



Leaflet | Data by © OpenStreetMap, under ODbL



# 3. Methodology

- 3.1 Definition of Sufficiency and InSufficiency
- 3.2 Isolation Forest Anomaly Detection



## 3.1 Definition of Sufficiency and InSufficiency

### **Sufficient Demand :**

a bigger average restaurant provider over the area

### **Insufficient Support:**

a smaller average restaurant business over the area



## 3.2 Isolation Forest Anomaly Detection

- Step (1) Define and Fit the model

```
IF_cols = ['Avg_ChineseR','Avg_AsianR','Avg_OtherR','Avg_AnyR']
```

```
model=IsolationForest( n_estimators=50, max_samples='auto', contamination=float(0.1),  
max_features=4 )
```

```
model.fit( df_avg[])
```



## 3.2 Isolation Forest Anomaly Detection

- Step (2) Attached Scores and Anomaly Column

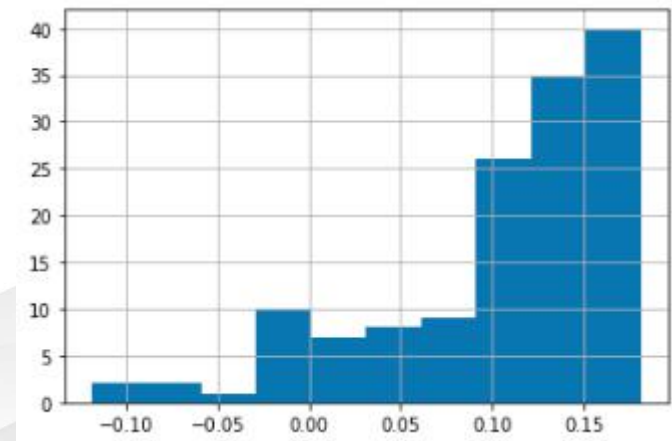
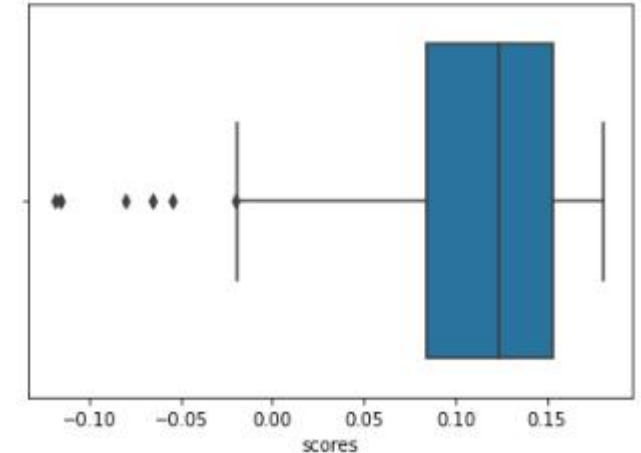
```
result_cols = ['AREA_ID', 'scores', 'anomaly', 'NAME', 'LATITUDE', 'LONGITUDE' ]  
df_avg['scores'] = model.decision_function(df_avg[IF_cols])  
df_avg['anomaly']=model.predict(df_avg[IF_cols])  
df_result = df_avg.merge( df )[result_cols]  
df_result.sort_values('scores', inplace=True)
```





## 3.2 Isolation Forest Anomaly Detection

	AREA_ID	scores	anomaly	NAME	LATITUDE	LONGITUDE
27	4649	-0.118717	-1	North St.James Town (74)	43.669623	-79.375247
37	4660	-0.115751	-1	Regent Park (72)	43.659992	-79.360509
2	4623	-0.079746	-1	Yonge-St.Clair (97)	43.687859	-79.397871
42	4665	-0.065879	-1	Rouge (131)	43.821201	-79.186343
40	4663	-0.055065	-1	Roncesvalles (86)	43.646123	-79.442992
...	...	...	...	...	...	...
24	4646	0.173633	1	Newtonbrook West (36)	43.785830	-79.431422
113	4739	0.174701	1	Forest Hill South (101)	43.694526	-79.414318
39	4662	0.177534	1	Rockcliffe-Smythe (111)	43.674790	-79.494420
41	4664	0.180706	1	Rosedale-Moore Park (98)	43.682820	-79.379669
9	4630	0.181015	1	Leaside-Bennington (56)	43.703797	-79.366072



# 4. Result

To open a new Chinese restaurant, we have two locations with potentially highest opportunities, marked as Red on above map:

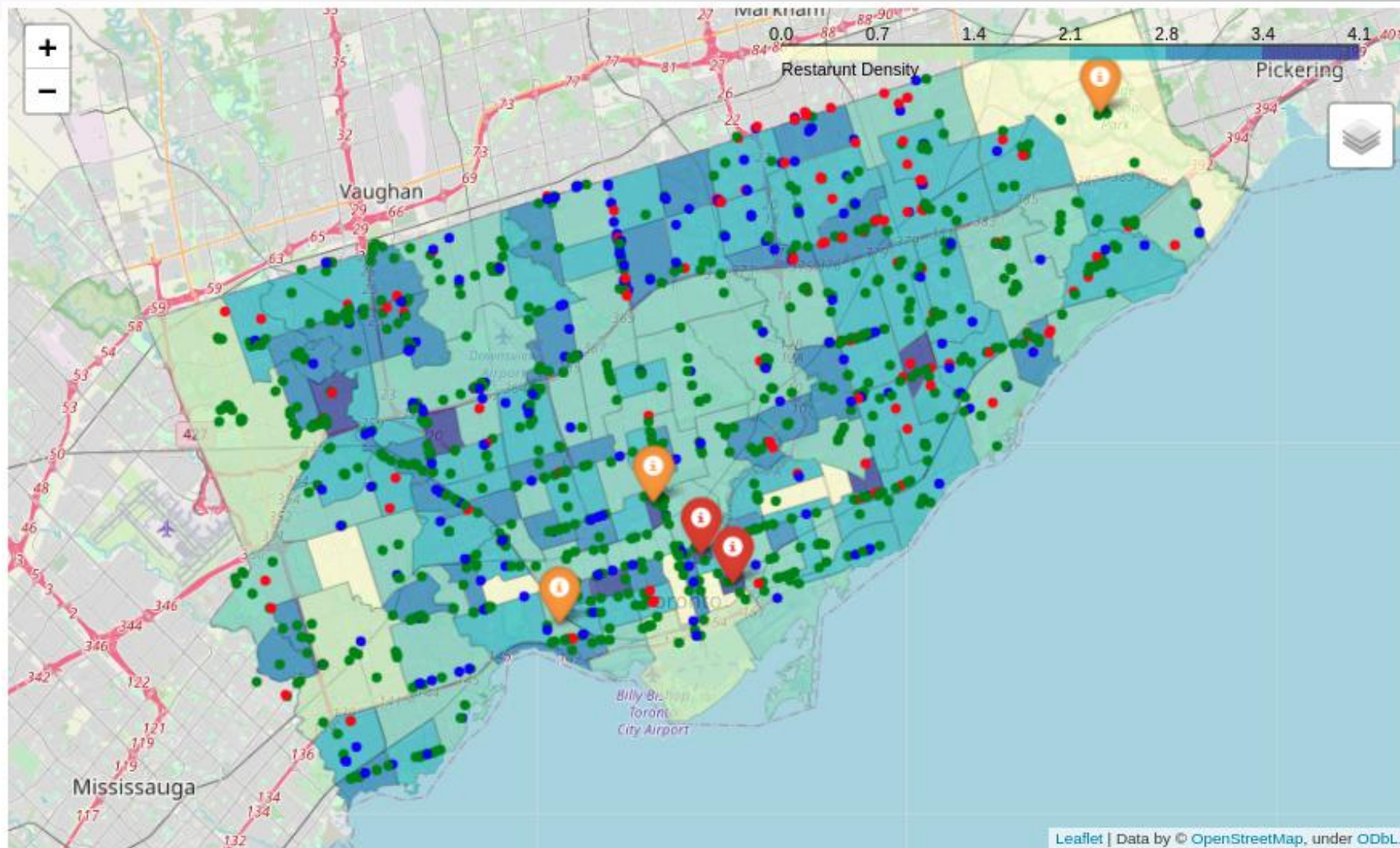
- North St.James Town
- Regent Park

Plus the other three locations may also have moderate opportunities, marked as Orange on above map:

- Yonge-St.Clair
- Rouge
- Roncesvalles



# 4. Result



# 5. Discussion

- We could add more features into the model, such as the rating of the venues, size of the business, etc.
- Also, it would be better if we could fetch more data from different data sources, along with FourQuares, it may help us to build a more accurate model.
- Plus, We could introduce other dimensional data like population, Demographics, income, etc., for this information also has an impact on the consuming market.



## 6. Conclusion

**Isolation Forest model** help us to find **outliers** in the restaurant business, and find out location-based significantly different from those majority of the other locations



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# Thank You!

This is Peer-graded Assignment for Course Applied Data Science Capstone, Week 4/5

