THE BATTLE OF NEIGHBOURHOODS

CITY OF TORONTO

1. Introduction

Problem

Which neighbourhoods have the similar features from where a family with young children can purchase a detached residential property? Who are interested?

Families with young children of school age who are looking for a residential property in City of Toronto.

Data Acquisition and Cleaning

- FSA postcodes of City of Toronto from Wikipedia
- Geospatial coordinates of each postcode



□ Venues in each
postcode from
Foursqaure

Limit top 10 venue categories

Coffee Shop	303
Café	169
Park	117
Pizza Place	113
Restaurant	102
Italian Restaurant	97
Bakery	94
Sandwich Place	78
Hotel	68
Bar	68

- Elementary and SecondarySchools in each postcode from Foursqaure
- School ratings from Fraser Institute

	Postcode	Latitude	Longitude	Avg Rating Elementary Schools	Avg Rating High Schools
0	M1B	43.806686	-79.194353	6.240000	5.933333
1	M1C	43.784535	-79.160497	6.950000	4.125000
2	M1E	43.763573	-79.188711	5.033333	5.033333
3	M1G	43.770992	-79.216917	4.075000	5.750000
4	M1H	43.773136	-79.239476	5.200000	3.800000

98	M9N	43.706876	-79.518188	6.500000	0.000000
99	M9P	43.696319	-79.532242	5.900000	7.800000
100	M9R	43.688905	-79.554724	5.900000	7.266667
101	M9V	43.739416	-79.588437	5.200000	7.966667
102	M9W	43.706748	-79.594054	3.850000	7.450000

103 rows x 5 columns

- Transportation services in each postcode from Foursquire
- High count signifies easy transportation access

	Postcode	Latitude	Longitude	Transportation Count
0	M1B	43.806686	-79.194353	3
1	M1C	43.784535	-79.160497	1
2	M1E	43.763573	-79.188711	5
3	M1G	43.770992	-79.216917	0
4	M1H	43.773136	-79.239476	1
***	1.0.00	•••	***	***
98	M9N	43.706876	-79.518188	0
99	М9Р	43.696319	-79.532242	1
100	M9R	43.688905	-79.554724	7
101	M9V	43.739416	-79.588437	3
102	M9W	43.706748	-79.594054	0

103 rows x 4 columns

2016 Census from Statistic Canada

Data	HTML headers attribute value
Population size	'L1000 geo1 total1'
Median age	'L2033 geo1 total1'
Average household size	'L3017 geo1 total1'
Without children in a census family	'L6003 geo1 total1'
With children in a census family	'L6004 geo1 total1'
Number of households	'L6001 geo1 total1'
Median total income	'L13001 geo1 total1'
Non-immigrants	'L18001 geo1 total1'
ImmigrantsCensus	'L18002 geo1 total1'
Non-permanent residents	'L18010 geo1 total1'

	Postcode	Census Population Size	Census Median Age	Census Average Household Size	Census Households With Children (%)	Census Number of Households	Census Median Family Income	Census Non Immigrants (%)	Census Immigrants (%)	Census Non Permanent Residents (%)
0	M1B	66108	38.2	3.3	80.10	20230	69126.0	38.21	59.92	1.87
1	M1C	35626	44.0	3.1	70.14	11275	109785.0	54.12	44.89	0.99
2	M1E	46943	42.2	2.7	72.57	17160	62047.0	50.74	47.79	1.47
3	M1G	29690	37.2	3.0	76.57	9765	54450.0	40.54	56.03	3.43
4	М1Н	24383	38.1	2.7	70.77	8985	58492.0	37.27	57.95	4.77
	***	1999	,,,,	***			***	3000		***
98	M9N	25074	39.0	2.4	73.68	10170	50545.0	49.56	47.74	2.71
99	M9P	20874	45.9	2.6	67.12	7905	73425.0	55.01	43.82	1.17
100	M9R	33743	40.6	2.7	70.89	12335	63032.0	47.34	50.69	1.97
101	M9V	55959	35.9	3.3	80.15	16805	59760.0	34.74	62.55	2.71
102	M9W	40684	38.3	2.9	74.47	13705	65826.0	44.18	52.79	3.02

103 rows x 10 columns

3. Methodology and Results

- Normalized data
- School ratings dived by 10
- Percentage divided by 100

	Postcode	Coffee Shop	Café	Park	Pizza Place	Restaurant	Italian Restaurant	Bakery	Sandwich Place	Bar	Hotel	Avg Rating Elementary Schools	Avg Rating High Schools	Transportation Count	Census Population Size
0	M1B	0.083333	0.0	0.000000	0.000000	0.0	0.0	0.000000	0.000000	0.0	0.0	0.624000	0.593333	0.061224	0.871023
1	M1C	0.000000	0.0	0.000000	0.000000	0.0	0.2	0.000000	0.000000	0.2	0.0	0.695000	0.412500	0.020408	0.469399
2	M1E	0.133333	0.0	0.000000	0.200000	0.0	0.0	0.000000	0.000000	0.0	0.0	0.503333	0.503333	0.102041	0.618509
3	M1G	0.333333	0.0	0.333333	0.000000	0.0	0.0	0.000000	0.000000	0.0	0.0	0.407500	0.575000	0.000000	0.391188
4	M1H	0.090909	0.0	0.000000	0.000000	0.0	0.0	0.090909	0.000000	0.0	0.0	0.520000	0.380000	0.020408	0.321264
***	2007					***	(500	866				866		776	142
98	M9N	0.000000	0.0	0.000000	0.166667	0.0	0.0	0.000000	0.000000	0.0	0.0	0.650000	0.000000	0.000000	0.330369
99	М9Р	0.100000	0.0	0.000000	0.100000	0.0	0.0	0.000000	0.100000	0.0	0.0	0.590000	0.780000	0.020408	0.275031
100	M9R	0.071429	0.0	0.000000	0.071429	0.0	0.0	0.000000	0.071429	0.0	0.0	0.590000	0.726667	0.142857	0.444589
101	M9V	0.076923	0.0	0.000000	0.230769	0.0	0.0	0.000000	0.076923	0.0	0.0	0.520000	0.796667	0.061224	0.737302
102	M9W	0.000000	0.0	0.000000	0.000000	0.0	0.0	0.000000	0.000000	0.0	0.0	0.385000	0.745000	0.000000	0.536042
103 r	ows × 23 c	columns													

3. Methodology and Results (cont.)

45069.875000 36.225000

22639.000000 41.911765

7154.750000 33.975000

42533,920000 40,116000

38.500000

7.224896 6.532054

7,443186 6,889511

5.737500 5.771250

6.014133 6.015467

K-Means clustering with K=7

Cluster label 0: 7
Cluster label 1: 18
Cluster label 2: 7
Cluster label 3: 8
Cluster label 4: 34
Cluster label 5: 4

Cluster label 6: 25

Cluster Labels	Avg Rating Elementary Schools	Avg Rating High Schools	Transportation Count	Census Population Size	Census Median Age	Census Average Household Size	Census Households With Children (%)	Census Number of Households	Census Median Family Income	Census Non Immigrants (%)	Census Immigrants (%)	Census Non Permanent Residents (%)
5.11	0.000000	6.909841	2.428571	24102.428571	42.142857	0.571400	63.782857	9640.000000	69550.285714	42.551429	E2 604006	3.844286
0	0.000000	0.909041	2,420071	24102.420071	42.142007	2.571429	03.702037	9040.000000	09000.200714	42.001429	53.604286	3.044200
1	6.426481	4.830992	3.888889	18951.388889	37.816667	2.511111	67.159444	7534.722222	56140.000000	43.737778	51.050000	5.212222
2	5.076531	5.841111	30.428571	6.428571	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000

1.887500

2.338235

1.625000

31.440000

15151.600000

54.820000

55.617500

61639.250000

39,465000

38.465000

56.274000

5.716250

2.217941

5.920000

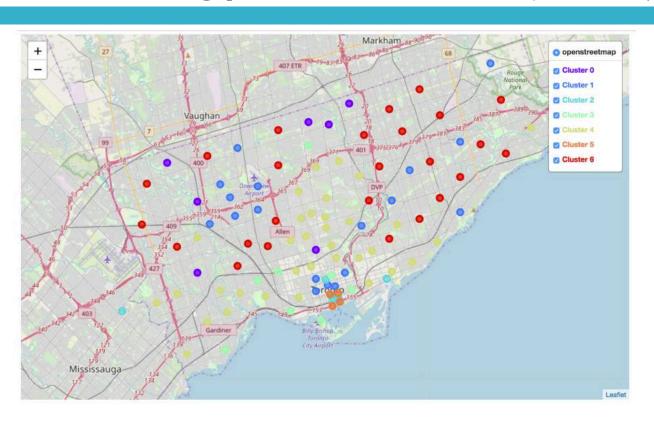
3.056400

3. Methodology and Results (cont.)

Averages of each cluster

	Avg Rating Elementary Schools	Avg Rating High Schools	Transportation Count	Census Population Size	Census Median Age	Census Average Household Size	Census Households With Children (%)	Census Number of Households	Census Median Family Income	Census Non Immigrants (%)	Census Immigrants (%)	Census Non Permanent Residents (%)
Cluster Labels												_
0	0.000000	6.909841	2.428571	24102.428571	42.142857	2.571429	63.782857	9640.000000	69550.285714	42.551429	53.604286	3.844286
1	6.426481	4.830992	3.888889	18951.388889	37.816667	2.511111	67.159444	7534.722222	56140.000000	43.737778	51.050000	5.212222
2	5.076531	5.841111	30.428571	6.428571	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
3	7.224896	6.532054	3.875000	45069.875000	36.225000	1.887500	47.798750	23288.125000	61639.250000	54.820000	39.465000	5.716250
4	7.443186	6.889511	2.500000	22639.000000	41.911765	2.338235	61.495588	9636.323529	88599.441176	65.060294	32.721765	2.217941
5	5.737500	5.771250	38.500000	7154.750000	33.975000	1.625000	31.440000	4322.500000	86254.750000	55.617500	38.465000	5.920000
6	6.014133	6.015467	3.040000	42533.920000	40.116000	2.776000	72.959600	15151.600000	60056.360000	40.668800	56.274000	3.056400

3. Methodology and Results (cont.)



4. Data Analysis

Cluster 0: Good Schools and Families with Children

- Top venues: Eateries and parks
- Good secondary schools but no elementary school rating found
- Many families with children (63.7%)
- Medium population density

Cluster 1: Good Elementary Schools and Young Families with Children

- Top venues: Eateries and parks
- Good elementary schools but poor secondary school ratings
- Young families in the neighbourhoods (Census Median Age < 40)
- Many families with children (67%)
- Low population density
- Look like these were new neighbourhoods in development

4. Data Analysis (cont.)

Cluster 2: Postcodes with insufficient census data

- There were no census data (zero most of the cases) in these postcode areas because the census population size was very small (less than 15)
- No characteristics could be determined

Cluster 3: Downtown Toronto, Average Schools, and High Population Density

- Top venues: Eateries (coffee, pizza, and restaurants) and hotels
- Household Size was small, at 1.89 persons per household
- Both Elementary and Secondary School Ratings were average to good
- High Population Density due to the highest Census Population Size with small Household Size, most likely high rise apartments

4. Data Analysis (cont.)

Cluster 4: Good Schools, High Family Income, and Non-Immigrants

- Top venues: Eateries and parks
- Good elementary and secondary school ratings
- High family income
- Predominantly non-immigrants (Census non-Immigrants = 65%)
- Older neighbourhoods located South of Highway 401

Cluster 5: Downtown Toronto, High Family Income, and Young Families

- Top venues: Eateries and hotels
- Schools were poor with both ratings below 6
- Household Size was small, at 1.62 persons per houshold, which was consistent with the low Family With Children (31%)
- High Census Median Family Income (over \$82k)
- Look like mostly young (Census Median Age = 34) professional couples without children lived in these neighbourhood

4. Data Analysis (cont.)

Cluster 6: Newer Neighbourhoods with High Population Density and Families with Children

- To venues: Eateries and parks
- Schools were average
- High population density with large Census Population Size (over 42k persons) and the highest Census Average Household Size (2.78 persons per household)
- 73% families with children, significantly higher than other clusters

Additional Observations in General about Toronto

- It seems elementary and secondary school ratings were poor in general, mostly 6 or below
- Non-immigrants and immigrants ratios was close to 50:50 in average
- Household size was small with average below 3

5. Discussion

Issues on Data Quality

- Foursquare venues were more on leisure places
- Did not have school ratings for all schools
- 2016 Census data were 3 years old

Difficulties on Data Acquisition

 Could not scrap and match communities from HouseSigma to get median pricing for residential properties

6. Conclusion

A family can determine which features are the most important and then search the neighbourhoods in the cluster, which has the features the family is looking for.

However, there are "human factors" that cannot include in "mathematical analysis" which sway the house purchasing decision. Examples are: nearness to other family members or friends, closeness to employment sites, ambient of the neighbourhood (such as country style), ravines or old large trees, etc.