Clustering of Neighbourhoods in Toronto to Aid Selection of Residence

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COGNITIVE CLASS.AI Coursera

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Introduction

Introduction

Problem statement: How can an individual find their next ideal neighbourhood of residence

Introduction

Objective: Segment and cluster all neighbourhoods in Toronto based on census and non-census data

Data acquisition and exploration

Data acquisition and exploration

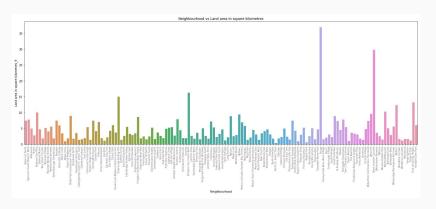
The data for this project is grouped into two classes:

- Census data. For example, population size, population density etc.
- · Non-Census data. For example, bus stations, cafes etc.

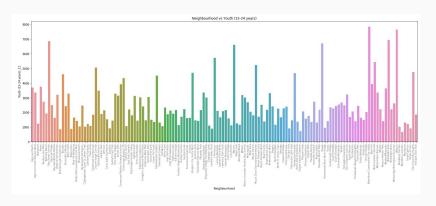
The census data was retrieved from the City of Toronto website. Below is sample of the census data:

| | je | Category | Topic | Data Source | Characteristic | City of Toronto | Agincourt North | Agincourt South- Malvern West | Alderwood | Arnex . | Willowdale West | Willowridge- Martingrove- Richview | Woburn | Woodbine Corridor | Woodbine- Lunsden | Wychwood | Yonge- Eglinton | Yonge- St.Clair | York University Heights | Yorkdale-Glen Park |
|------|------|------------------------------|------------------------------|------------------------------------|------------------------------------|--------------------|--------------------|----------------------------------|-------------------|-------------------|--------------------|--|--------|----------------------|----------------------|-------------------|--------------------|--------------------|-------------------------------|---------------------------|
| 0 | 1 | Neighbourhood Information | Neighbourhood Information | City of Toronto | Neighbourhood Number | NaN | 129 | 128 | 20 | 95 . | . 37 | 7 | 137 | 64 | 60 | 94 | 100 | 97 | 27 | 31 |
| 1 | 2 | Neighbourhood Information | Neighbourhood Information | City of Toronto | TBN82020 Designation | NaN | No Designation | No Designation | No Designation | No Designation | No Designation | No Designation | NA | No Designation | No Designation | No Designation | No Designation | No Designation | NIA | Emerging Neighbourhood |
| 2 | 3 | Population | Population and dwellings | Census Profile 98- 316-X2016001 | Population, 2016 | 2,731,571 | 29,113 | 23,757 | 12,054 | 30,526 | 16,936 | 22,156 | 53,495 | 12,541 | 7,865 | 14,349 | 11,817 | 12,528 | 27,593 | 14,804 |
| 3 | 4 | Population | Population and dwellings | Census Profile 98- 316-X2016001 | Population, 2011 | 2,615,060 | 30,279 | 21,988 | 11,904 | 29,177 | 15,004 | 21,343 | 53,350 | 11,703 | 7,826 | 13,986 | 10,578 | 11,652 | 27,713 | 14,687 |
| 4 | 5 | Population | Population and dwellings | Census Profile 98- 316-X2016001 | Population Change 2011- 2016 | 4.50% | -3.90% | 8.00% | 1.30% | 4.60% . | 12.90% | 3.80% | 0.30% | 7.20% | 0.50% | 2.60% | 11.70% | 7.50% | -0.40% | 0.80% |
| 5 10 | ws x | 146 columns | | | | | | | | | | | | | | | | | | |

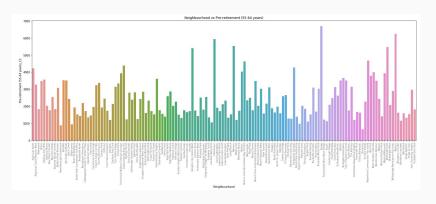
Below is a plot of the Neighbourhood vs the Neighbourhood Land Area:



Below is a plot of the Neighbourhood vs the population of Youths (15-24years):



Below is a plot of the Neighbourhood vs the population of Pre-retirement residents (55-64years):



Non-Census data

The Non-census data was retrieved leveraging the Foursquare and Google-Maps API. Below is a sample of the Non-census data:

| | Neighbourhood | Afghan Restaurant | Airport Service | American Restaurant | Antique Shop | Argentinian Restaurant | Art Gallery | Art Museum | Arts & Crafts Store | Asian Restaurant | Video Store | Vietnamese Restaurant | Volleyball Court | Warehouse Store | Whisky Ber | Wine Bar | Wine Shop | Wings Joint | Women's Store | Yoga Studio |
|----------------------|-----------------|----------------------|--------------------|------------------------|-----------------|---------------------------|----------------|---------------|------------------------|---------------------|----------------|--------------------------|---------------------|--------------------|---------------|-------------|--------------|----------------|------------------|----------------|
| 0 | Agincourt North | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 1 | Agincourt North | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 2 | Agincourt North | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 3 | Agincourt North | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 4 | Agincourt North | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 5 rows × 269 columns | | | | | | | | | | | | | | | | | | | | |

Merged data

Leveraging Pandas concat function, the census and non-census data are merged. Below is a sample of the merged data:

| | Neighbourhood | Afghan Restaurant | | American Restaurant | | Argentinian Restaurant | Art Gallery | Art Museum | Arts & Crafts Store | Asian Restaurant | External migrants_2374 | Total - Mobility status 5 years ago - 25% sample data_2375 | Non- movers_2376 | Movers_2377 | Non- migrants_2378 | Migrants_2379 | Internal migrants_2380 | Intraprovincial migrants_2381 | Interprovincial migrants_2382 | External migrants_2383 |
|----|----------------------------------|----------------------|-----|------------------------|-----|---------------------------|----------------|---------------|---------------------------|---------------------|---------------------------|---|---------------------|-------------|-----------------------|---------------|---------------------------|----------------------------------|----------------------------------|---------------------------|
| 0 | Agincourt North | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.191104 | 0.373337 | 0.557659 | 0.149203 | 0.201777 | 0.109452 | 0.040462 | 0.045435 | 0.028708 | 0.213190 |
| i | Agincourt South- Malvern West | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.153213 | 0.283503 | 0.371889 | 0.152969 | 0.210152 | 0.108431 | 0.046547 | 0.048473 | 0.049043 | 0.201943 |
| 2 | Alderwood | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.014827 | 0.092965 | 0.185068 | 0.024279 | 0.037056 | 0.016904 | 0.028293 | 0.036660 | 0.013158 | 0.005112 |
| 3 | Annex | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.266995 | 0.377250 | 0.351385 | 0.288841 | 0.348731 | 0.242218 | 0.226042 | 0.200815 | 0.309939 | 0.231595 |
| 4 | Banbury-Don Mills | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.116969 | 0.346117 | 0.467753 | 0.172347 | 0.254315 | 0.108225 | 0.072406 | 0.083503 | 0.049043 | 0.157454 |
| ro | ws × 2594 colur | mns | | | | | | | | | | | | | | | | | | |

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Merged data

Below is map highlighting all the neighbourhoods in Toronto:



Methodology

Clustering algorithms

The problem statement is a unsupervised machine learning problem. It requires the use of clustering algorithms. Two clustering algorithms are used in this project:

- · K-means clustering
- HDBSCAN

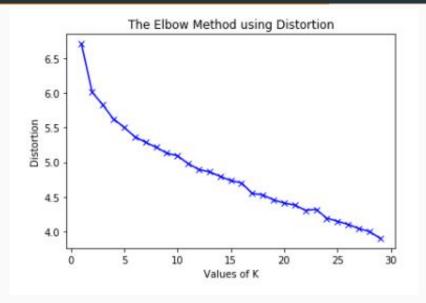
Clustering algorithms

K-means clustering data assumption:

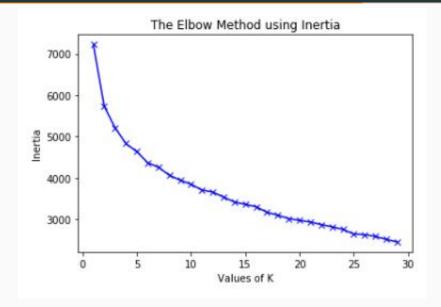
- Round and spherical clusters
- · Equally sized and dense clusters
- · Clusters with high density at the center
- · Absence of noise and outliers in the data

Results

K-Means: Elbow Method (Distortion)



K-Means: Elbow Method (Inertia)



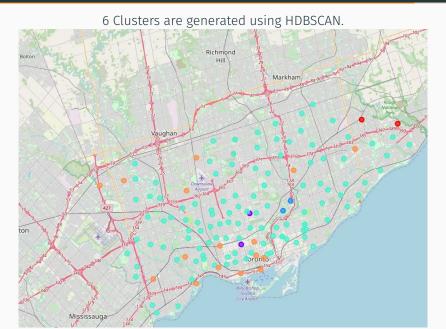
K-Means: Map



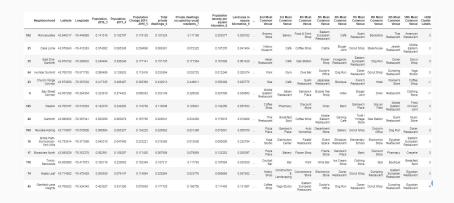
The K-Means assumptions are not met.

It's not clear the optimal number of clusters from the elbow method.

HDBSCAN: Map



Sample Neighbourhoods in their clusters



Conclusion

Conclusion

- HDBSCAN is the suitable clustering algorithm for this project.
- The neighbourhoods are grouped into 6 clusters.

The End Thank you!