## Analysis on break and enter crime in Toronto from 2014 to 2019

### Introduction

The aim of this project is to find a safe and secure location for living or opening a business in Toronto, Canada. Specifically, this report will be targeted to the break and enter crime situation over the past 5 years in different neighbourhood in Toronto.

We will make use of our data science tools to analyse data and focus on the trend of each borough from 2014 to 2019, and will focus on the 2019 data specifically to choose the safest and most dangerous boroughs by analysing crime data and short listing the neighbourhood.

#### Data

Based on definition of our problem, factors that will influence our decision are:

finding the most dangerous borough based on crime statistics in 2019

finding the trend of beak and enter crime for each borough

We will be using the geographical coordinates of Toronto to plot neighbourhoods in a borough that is dangerous and in the city's vicinity, and finally cluster our neighborhoods and present our findings.

Following data sources will be needed to extract/generate the required information:

Part 1: Using a real world data set from Tronto Police Service website: A dataset consisting of the crime statistics of each Neighbourhood in Toronto along with location, recorded year, premise type.

https://data.torontopolice.on.ca/datasets/break-and-enter-2014-to-2019/data https://prod-hub-indexer.s3.amazonaws.com/files/d9b3dd6402454c379ba57994230aabea/0/full/3857/d9b3dd6402454c379ba57994230aabea 0 full 3857.csv

Part 2: Gathering additional information of the list of officially categorized boroughs in Toronto: Borough information will be used to map the existing data where each neighbourhood can be assigned with the right borough.

Part 3: Creating a new consolidated dataset of the Neighborhoods, along with their boroughs, crime data and the respective Neighbourhood's co-ordinates.: This data will be fetched using OpenCage Geocoder to find the safest and most dangerous borough and explore the neighbourhood by plotting it on maps using Folium and perform exploratory data analysis.

## Methodology

Categorized the methodology section into two parts:

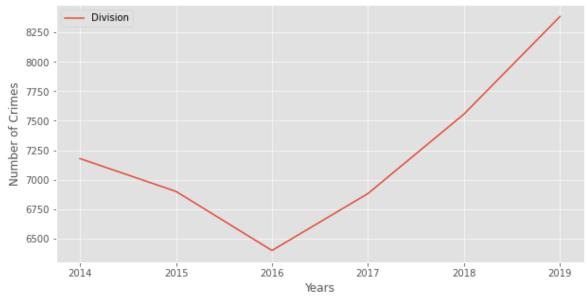
Exploratory Data Analysis: Visualise the crime reports in different Toronto district to idenity the 10 most dangerous area. Also to see the trend of each year

Modelling: To help stakeholders choose the right neighborhood within a borough we will be clustering similar neighborhoods using K - means clustering which is a form of unsupervised machine learning algorithm that clusters data based on predefined cluster size.

We will use K-Means clustering to address this problem so as to group data based on existing venues which will help in the decision making process.

# 1. see the trend for the break and enter from 2014 to 2019





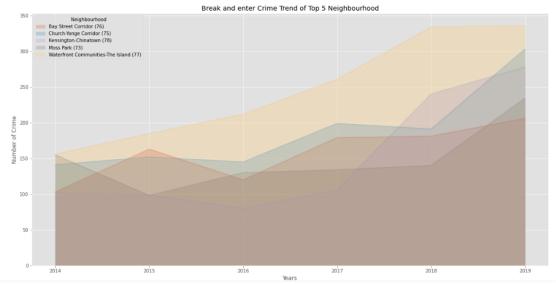
we can observe from the chart, the break and enter crime decrease to the bottom in 2016, and from then on, it has increased sharpely.

2. Let's continue to see the trend for the 5 most dangerous areas trend over these 5 years.

| Waterfront Communities-The Island (77)   | 1484  |
|--|-------|
| Church-Yonge Corridor (75)               | 1131  |
| Bay Street Corridor (76)                 | 952   |
| Kensington-Chinatown (78)                | 905   |
| Moss Park (73)                           | 891   |
|  |       |
| Caledonia-Fairbank (109)                 | 87    |
| Guildwood (140)                          | 87    |
| Maple Leaf (29)                          | 76    |
| Lambton Baby Point (114)                 | 75    |
| Elms-Old Rexdale (5)                     | 63    |
| Name: Neighbourhood, Length: 140, dtype: | int64 |
|  |       |

they are Waterfront Communities-The Island (77), Church-Yonge Corridor (75), Bay Street Corridor (76), Kensington-Chinatown (78), and Moss Park (73):

| Neighbourhood |              | Bay Street Corridor (76) | Church-Yonge Corridor (75) | Kensington-Chinatown (78) | Moss Park (73) | Waterfront Communities-The Island (77) |  |
|---------------|--------------|--------------------------|----------------------------|---------------------------|----------------|--|--|
|               | reportedyear |                          |                            |                           |                |  |  |
|               | 2014         | 103                      | 141                        | 102                       | 155            | 156                                    |  |
|               | 2015         | 163                      | 152                        | 99                        | 98             | 185                                    |  |
|               | 2016         | 120                      | 145                        | 80                        | 130            | 212                                    |  |
|               | 2017         | 179                      | 199                        | 106                       | 134            | 261                                    |  |
|               | 2018         | 181                      | 191                        | 240                       | 140            | 334                                    |  |
|               | 2019         | 206                      | 303                        | 278                       | 234            | 336                                    |  |

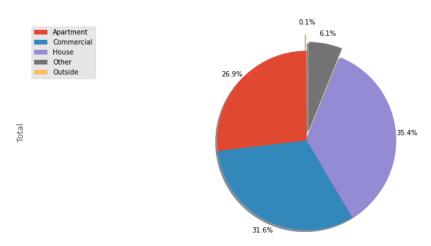


from the chart, all five areas the crime are on the trend, and the water front community (77) are higher than the other area.

# 3. The premise type analysis

| reportedyear | 2014   | 2015   | 2016   | 2017   | 2018   | 2019   | Total   |
|--------------|--------|--------|--------|--------|--------|--------|---------|
| premisetype  |        |        |        |        |        |        |         |
| Apartment    | 1865.0 | 1982.0 | 1599.0 | 1818.0 | 1986.0 | 2398.0 | 11648.0 |
| Commercial   | 1916.0 | 1933.0 | 1900.0 | 2118.0 | 2552.0 | 3263.0 | 13682.0 |
| House        | 2941.0 | 2637.0 | 2495.0 | 2575.0 | 2505.0 | 2169.0 | 15322.0 |
| Other        | 457.0  | 346.0  | 402.0  | 369.0  | 511.0  | 541.0  | 2626.0  |
| Outside      | NaN    | 2.0    | 2.0    | 2.0    | 3.0    | 15.0   | 24.0    |

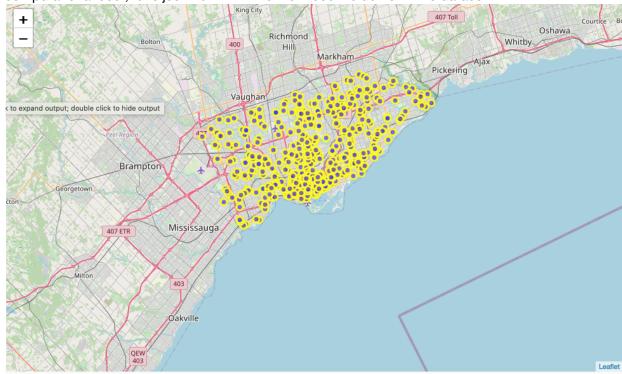
### Break and Enter Crime by Type [2014 - 2019]



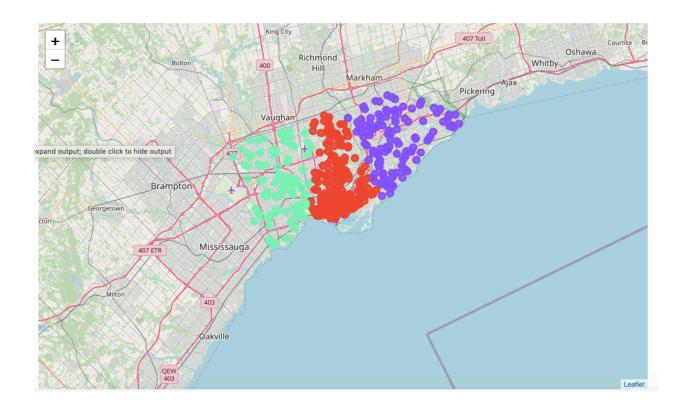
Houses and commercial estate take up most of the break and Enter Crime type.

# 4. Visualization and k-means modeling

The dataframe consists of 8386 crimes, which took place in the year 2019. In order to reduce computational cost, let's just work with the first 1000 incidents in this dataset.



We use k=3 to divide them into 3 clusters:



## **Result and Discussion**

The objective of the problem was to help stakeholders identify the most dangerous borough in Toronto to purchase a house or commercial estate. This has been achieved by first making use of Toronto crime data. We achieved this by grouping the neighborhoods into clusters to assist the police or the government to decide different area to govern by providing them with relavent data.

### Conclusion

We have explored the crime data to understand break and enter crimes in all neighborhoods of Toronto and later categorized them into different boroughs, this helped us group the neighborhoods into boroughs and choose the dangerous borough first. We can see from the chart that the crime trend is on the rise, and the break and enter crime into house is the most common type. The most dangerous areas are concentrated in the downtown area. The whole crime map can be divided into 3 clusters to let police or government to manage.