

# Toronto Major Crime Trends and Analysis

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This paper investigates major crime trends in Toronto from 2014-2024, using data sourced from OpenDataToronto. Of particular interest, the data shows that assault, motor vehicle theft and the total number of major crimes have been steadily on the rise throughout the 10 years this data has been recorded. This research is significant as it provides insights into crime in Canada's largest city; it aims to help build an understanding of the nature of Toronto's crime scene.

## 1 Introduction

Crime data analysis plays a crucial role in understanding social behavior and informing public policy. The city of Toronto, like many major urban centers, faces significant challenges with crime, and accurate data collection is vital to monitor trends, allocate resources effectively, and implement preventative strategies. While many countries and cities publish crime data, not all provide detailed datasets like the "Major Crime Indicators" dataset made available by Toronto Police Services through OpenDataToronto. This paper seeks to analyze crime patterns in Toronto from 2014 to 2024, using this comprehensive dataset to uncover important trends and insights.

Despite the availability of this dataset, there are limitations to its accuracy and coverage. Sexual assaults are notably absent, and certain crimes may be underreported or miscategorized. Nevertheless, the data provides a valuable glimpse into the most commonly reported crimes, allowing for the identification of trends and emerging issues. These patterns are crucial in providing recommendations for crime prevention strategies. A detailed analysis of how crime patterns evolved over time fills a gap in the current understanding of urban crime in Toronto.

This paper uses Python (Van Rossum and Drake 2009), a widely adopted tool for data analysis, visualization, and statistical investigation, alongside libraries like Matplotlib (Hunter 2007) and Seaborn (Waskom 2021), to visualize and analyze trends within the data. The paper provides insights into the types of crimes most frequently reported, changes in crime rates over time, and how these trends differ based on location and time of year. In particular, we will focus

on understanding the rise of assaults, motor vehicle thefts, and break-and-enters, and explore their geographical and temporal distribution.

The following sections will delve into the structure of the dataset, the methodology used for cleaning and analyzing the data, the resulting visualizations of major crime trends, and a discussion on the implications of these trends for public policy and law enforcement strategies. By the end of this paper, we aim to offer a clearer understanding of crime dynamics in Toronto.

Code and data supporting this analysis is available.<sup>1</sup>

## 2 Data

The data used in this paper was gathered from OpenDataToronto (Gelfand 2022). OpenDataToronto is a publicly funded data-gathering service that collects and publishes data about Toronto and the GTA. The specific dataset used in this paper is called “Major Crime Indicators”, and was published by Toronto Police Services. The data includes “all Major Crime Indicator offences reported by date” In order to perform the analysis, the programming language for data analysis, visualization and statistical investigation Python was used (Van Rossum and Drake 2009), along with the Python packages Matplotlib.pyplot (Hunter 2007), Seaborn (Waskom 2021), Pandas (team 2020), Requests(Reitz and Team 2024), Pytest (Krekel and team 2024), IO (Foundation 2024b), Random (Foundation 2024a) and Pathlib (Foundation 2019).

### 2.1 The Dataset

As mentioned, the dataset used in this paper is the “Major Crime Indicators” from OpenDataToronto. The dataset includes around 400,000 samples each with 26 observations, including the report date, the occurrence date, the neighbourhood that the crime took place in, where the crime took place (whether it was in an apartment, a restaurant, outside, etc), and many more. No other datasets were considered because no other department has access to this information. There are a few interesting things to note with the dataset. The first is that sexual assaults were not included, it is not explained why this is done but it is an important thing to note when looking through the paper. Secondly, there are instances in which a crime has several instances of data if it was the case that someone broke into a house (B&E) and proceeded to assault the residences, this may be recorded as two separate instances (ie. one for B&E and one for assault). Lastly, the dataset does not include occurrences that police have investigated and determined did not occur, nor was it attempted. This may result in inaccurate reporting, and police methods are often wrong so it is highly likely that the data is not flawless and may leave some crimes out.

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<sup>1</sup>[https://github.com/AlexanderG123/toronto\\_major\\_crime](https://github.com/AlexanderG123/toronto_major_crime)

With respect to the cleaning of the dataset, I removed quite a few redundant columns (ie. there were 7 columns that had data on the occurrence date which seemed unnecessary). Also, I eliminated police-specific data points, like the ID of the crime which we have no use for in this analysis.

## **2.2 Variables of Interest**

The cleaned data includes four variables. The date of the offence, the type of offence, the neighbourhood the offence took place in and the location type (ie. apartment, etc). The date of the offence needs no explanation, but it is important to note that although this data started being recorded in 2014, some people reported crimes back as far as 1966, because this skews certain data points, all the analysis is done for crimes after 2014 (I left the older data in the cleaned dataset so people can examine it if they choose). The type of offence is also relatively straight forward, one aspect of it that is important is they do distinguish between different types of crimes. For example, assault and assault with a weapon are two different types of offences, and so is B&E and B&E with intent. The neighbourhood aspect is rather unambiguous, these are the 158 social planning neighbourhoods that Toronto is broken up into, and these are for almost used for all relevant datasets. And the location type is also straightforward.

## **2.3 Measurement**

The measurement of this data is not particularly complex and does not have too much room for error. The event takes place and then is reported to the police, the police categorize each instance, using the reported date, time, location, neighbourhood, and offence and this information is added to the dataset. The reporting method is somewhat flawed, it all depends on what the victim claims happened, it is certain that some details (like the hour the incident took place) are not completely accurate, as a lot of these crimes were reported quite long after the fact. There is also the police categorization of these crimes, for example, what distinguishes assault from assault with a weapon, and what is considered a weapon? These can create uncertainty and measurement errors in the data, but ultimately, because of how broad these categorizations are and their legal definitions, it should not detract from the following analysis.

## **3 Data Visualization and Analysis**

Our analysis explores the trends in major crime indicators in the GTA from 2014 to 2024.

The Figure 1 shows the overall breakdown of the top 20 types of major crimes committed over the study.

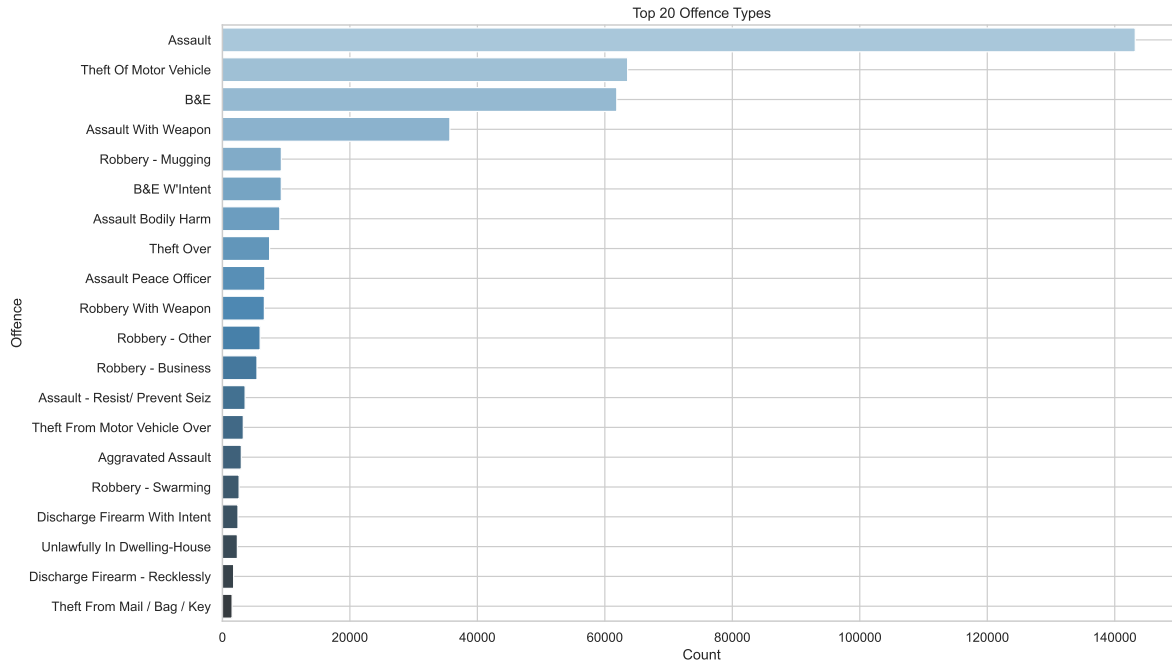


Figure 1

As you can see from the data, far and away the most common type of major crime is Assault with 144101 occurrences from 2014-2024. There were 63633 Motor Vehicle thefts and 61954 reported B&E's. The graph only shows the top 20 most common major crimes committed over this period, but the data shows that there were reports of 657 occurrences of "Administering Noxious Thing" and 1 instance of "Hoax Terrorism Causing Bodily Harm". The overwhelming amount of assaults relative to all other major crime indicators is particularly notable.

Figure 2 shows the trends in the top 5 major crimes from 2014 to 2024

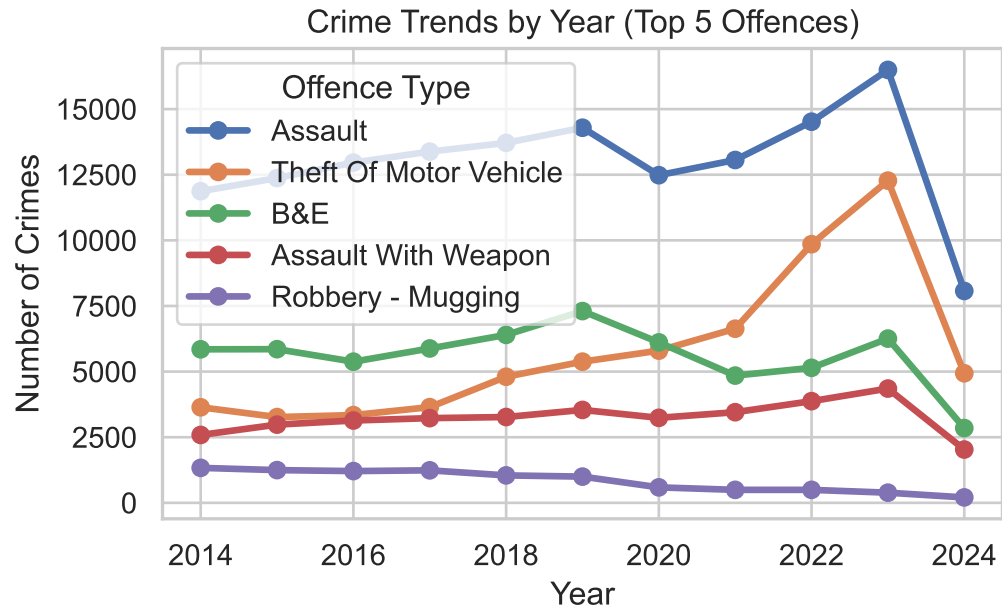


Figure 2

Figure 2 is somewhat misleading (the next graph Figure 3 does a good job of highlighting the issue with the above), but it does a good job of providing insights into the trends of these top 5 major crime indicators. The primary issue is that the 2024 data is not complete yet, the data is only up to date as of June 2024, this means that the down tick as of 2024 is not because of a decrease in crime but instead is because of a lack of data. With that explained, clearly the trend in assaults and theft of motor vehicles has been substantially increasing since the data started being collected. Whereas there has been some progress made with a slight decline in B&Es as well as Robberies. This graph shows how impactful covid was in reducing violent crime which is interesting.

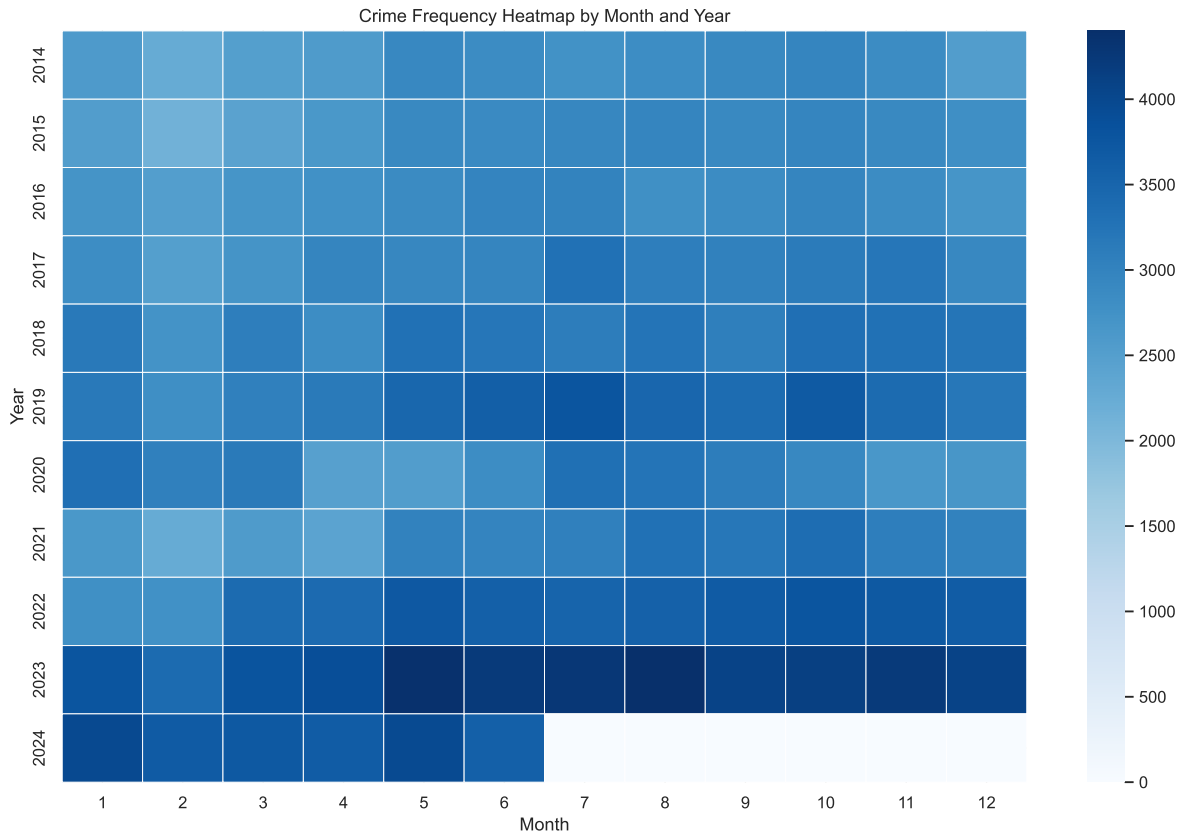


Figure 3

Figure 3 does a better job of showing the overall increasing trend in violence. The darker squares are months that have had more crimes. As you can see the last six months of 2024 have no data so there are zero reported crimes which makes 2024 look like there has been less crime, but if you look at a month by month basis, it becomes clear that major crime is on the rise.

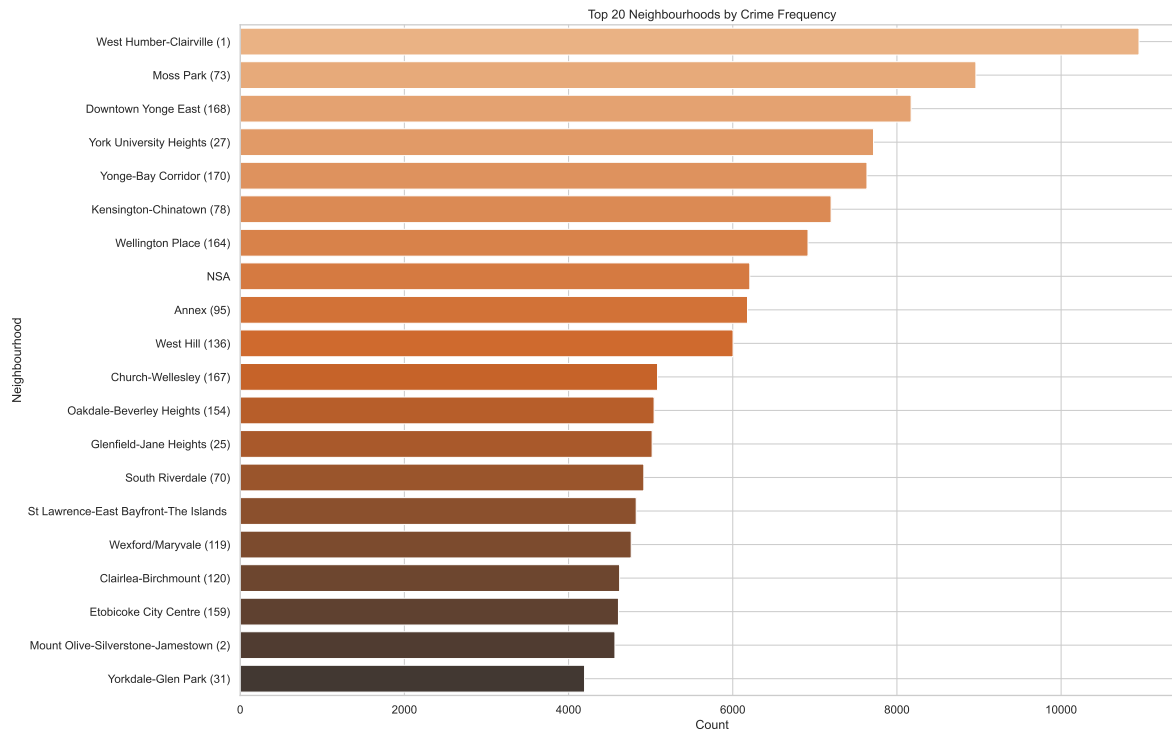


Figure 4

Figure 4 reveals the 20 neighbourhoods with the highest crime rates in the GTA. Understanding which neighbourhoods are most affected by crime is critical for local law enforcement and policymakers to develop targeted interventions. The frequency of major crimes is notably concentrated in certain areas, suggesting the need for enhanced security measures and community engagement in those neighbourhoods.

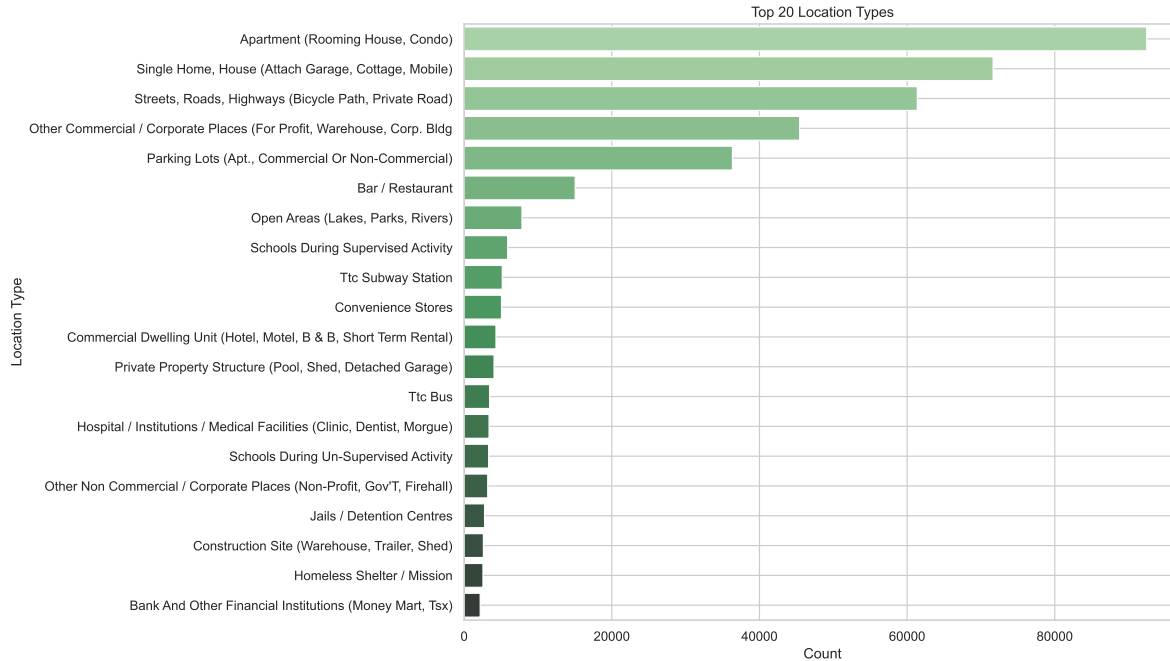


Figure 5

Figure 5 highlights the top 20 types of locations where major crimes were committed. Unsurprisingly, homes and apartments top the list. This raises an important question: Are these locations more prone to crime due to a higher likelihood of residents being present, or do people spend more time at home and thus report crimes there more frequently? Further analysis would be needed to explore this in greater detail.

Each of the figures provided offers a distinct view of crime trends and patterns across Toronto, with a focus on different aspects such as offence types, neighbourhoods, and locations. These insights can be pivotal in addressing crime prevention and policy formulation moving forward.

## Conclusion

This study delved into the “Major Crime Indicators” dataset from OpenDataToronto, offering a comprehensive analysis of crime trends in Toronto between 2014 and 2024. The findings highlight persistent and growing challenges for public safety, with assaults being the most frequently reported major crime, followed by motor vehicle thefts and break-and-enters. The rise in violent crimes, particularly assaults, underscores an urgent need for targeted law enforcement and community-based interventions to address these issues.



The spatial analysis of crime further highlights certain Toronto neighborhoods as more vulnerable to specific types of crimes. This geographic concentration emphasizes the necessity of localized, context-specific crime prevention strategies. Insights from this analysis provide a foundation for policymakers to enhance resource allocation, law enforcement presence, and community engagement in high-crime areas.

Despite the limitations of the dataset—including the absence of sexual assault data and potential misclassification of incidents—this study offers valuable insights that can inform public policy, law enforcement strategies, and future research. The exclusion of sexual assaults, in particular, raises important questions about transparency in crime reporting and data completeness. Future work should push for more comprehensive data collection to include under-reported or excluded categories of crime.

Ultimately, the findings of this paper illustrate the importance of large-scale data analysis in understanding urban crime dynamics. By leveraging powerful tools like Python and its data analysis libraries (Pandas, Matplotlib, Seaborn, and more), we can unlock patterns that were previously inaccessible or unnoticed. Moving forward, it is crucial that law enforcement agencies and policymakers adopt data-driven approaches to address the complexities of crime in Toronto. Continuous monitoring, public transparency, and community collaboration are essential to reducing crime rates and enhancing safety across the city.

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