Decoding Crime Patterns: Analysis of Torontos Crime Dynamics

Yuyuan Kang

Executive Summary

To gain an understanding of the activity patterns, in Toronto, our project utilized data visualization techniques to. Interpret the Major Crime Indicators (MCI) of the city. The main goal of this initiative was to offer insights on safety, for groups, including policymakers, law enforcement agencies, community organizations and the public.

Quick Link

Link to the original dataset: Major Crime Indicator Open Data

Link to the visualization: Shiny Application
Link to the GitHub Repo: Crimes in Toronto

Current Context

Urban areas frequently experience issues that contribute to trends in crime and safety challenges. The nature and intensity of criminal activities significantly influence a citys welfare, its urban development strategies, and political decision-making. Fluctuations in crime statistics can affect real estate values, commercial investments, and public perceptions of urban safety. Weisburd and associates highlight the critical role of specific small areas, or "micro places," in their outsized contribution to overall city crime rates (Weisburd, Groff, & Yang, 2012).

The city of Toronto, mirroring the complexities of other large metropolitan areas, faces the ongoing challenge of ensuring public safety and effectively managing crime. As Canadas most populous city, it serves as a microcosm of diverse neighborhoods, each with its unique security dynamics. As data enthusiasts and civic planners, we will turn to the numbers to paint a picture of safety across this urban landscape. We aim to create a user-accessible platform that enables citizens, policymakers, and researchers to comprehend and investigate crime dynamics in Toronto ultimately fostering a safer and more knowledgeable community.

Intended Audience

The project is designed to cater to a diverse audience, each with their unique interests and needs in urban safety and data utilization. The primary audience segments include:

- Law Enforcement Agencies: Police departments and other law enforcement can use the
 insights from the project to allocate resources, anticipate crime trends, and plan effective
 interventions.
- Academics and Researchers: Individuals doing academic research focusing on criminology, urban studies, sociology, or data visualization can utilize the findings as a base for further study, teaching, or to augment their research.
- **Public and Residents**: The public, especially residents, can use the result to gain a better understanding of the safety landscape of their environment and make wise decisions about their personal safety and community involvement.

Key Insights

This project focuses on a serious of pressing concerns central to the well-being of Toronto's communities, including:

- **Neighborhood Safety**: Identifying which city areas or neighborhoods are perceived as safer or more perilous based on reported crime incidents.
- **Time-Specific Trends**: Determining the months and days marked by elevated crime rates, is essential for strategic planning of law enforcement efforts and public awareness campaigns.
- **Premises at Risk**: Evaluating which types of premises are more susceptible to offenses, providing insights for property owners and renters on security measures.
- **Residential Decision-Making**: Guiding potential residents on safer areas to live by correlating crime data with residential zones.
- Law Enforcement Resource Deployment: Using the findings to recommend areas that could benefit from increased police presence or community-based crime prevention initiatives.

The Bedrock of Our Crime Analysis in Toronto

To embark on our journey of understanding Toronto's crime patterns, we meticulously sourced data that serves as the cornerstone of our project. Here, we outline the specifics of the data underpinning our analysis.

Data Sourcing and Integrity

Our expedition starts from the Major Crime Indicators (MCI) data, which has been diligently compiled by the Toronto Police Service and made readily available to the public on their open data portal. This extensive collection of information serves as our starting point offering a glimpse into the city reported crime through channels.

The data is published under the Open Government License - Toronto, which allows for the free use, modification, and sharing of the data, provided that the Toronto Police Service is acknowledged as the source. It grants us the freedom to explore, analyze and share our discoveries while acknowledging the meticulousness of the Toronto Police Service in documenting these incidents.

While the MCI data provides a high-resolution image of crime in Toronto, the granularity comes with a responsibility to uphold privacy. The data is prudently anonymized, with no trails leading back to individuals. Yet, we tread carefully, cognizant that even anonymized data can cast long shadows when it comes to privacy.

Collection and Bias

Collection Methods and Bias

The MCI data is collected by the Toronto Police Service through their operational activities. Each reported crime is logged into their system with details such as the type of crime, location, and time of occurrence. This administrative data collection method ensures a comprehensive log of reported criminal activity. To be notice, the data applies two different kinds of neighborhood divisions (neighborhood 140 and neighborhood 158). We choose 140 as our standard in our project.

As with any dataset dependent on reporting, there is potential for bias. Certain crimes may be underreported due to victim reluctance or lack of witness engagement, while others might be overrepresented due to active enforcement in specific areas. Sampling bias is inherently limited, as the dataset aims to capture all reported incidents, but it is essential to acknowledge these potential discrepancies. On the first day of each month, there is a notable surge showing in the number of reports. This may be because uncertain data is entered for convenience.

Descriptive Statistics

This dataset includes all Major Crime Indicators (MCI) occurrences by reported date and related offenses from 2014 to 2022, which contains more than 300,000 cases in the dataset. Therefore, we choose to focus on only crimes in 2018, which contains 36,406 cases. The data includes incidents, categorized into key crime types such as assault, auto theft, break and enter, robbery, and theft over. The main features we will focus on include the occurrence of the month, occurrence of date, MCI category, premises type, and division of neighborhoods. The distribution across these categories varies, with some crimes being more prevalent than others, and might be subject to seasonal and temporal variations.

The Tools for Crime Analysis

To transform raw data into a compelling story of Torontos crime landscape, weve harnessed a suite of packages in R. Each package plays a pivotal role in shaping our analysis and the interactive experience we aim to provide. Here's a rundown of the key packages:

- sf (Simple Features): The sf package plays a role in managing data. Using sf, we can easily handle geojson files by reading and processing them. It forms the foundation of our mapping enabling us to represent the neighborhoods of Toronto along with their corresponding crime data.
- tmap: When it comes to creating maps, our preferred tool is tmap. It seamlessly integrates with sf allowing us to transform data into captivating maps that not only show the locations of crimes but also convey the intensity and distribution of different crime types throughout Toronto.
- plotly: To add a touch to our visualizations we rely on plotly for heatmaps. Plotly enhances our charts by providing features such as zooming hovering and exploring information, about each crime category across types of premises. With Capabilities, we can turn charts into engaging experiences that actively involve users.
- shiny: our chosen medium for delivering an interactive user experience. It allows us to build web applications directly from R, turning static graphs into dynamic visual stories. With Shiny, users can manipulate variables, filter results, and watch the crime narrative unfold in real time.

A Summary of Our Analytical Journey

Our analytical expedition through Torontos crime landscape involved an intricate process of data visualization and exploration. Here's a summary of the key analytical steps undertaken:

Geospatial Analysis: Mapping Crime Across Neighborhoods

Using R's sf and tmap packages, we first embarked on a geospatial journey, plotting the citys

crime data onto the rich tapestry of Torontos neighborhoods. This step allowed us to visualize spatial patterns and identify which areas experience higher incidences of crime, providing a clear visual representation of safety disparities across the city. Additionally, we changed the base map to a more specific map which allows audiences to zoom in to the specific district.

Temporal Trends: Understanding When Crime Occurs

Next, we delved into the temporal domain, analyzing crime data across different time frames. We charted monthly trends to pinpoint when crime rates peaked and troughed throughout the year, uncovering patterns and outlier events. This analysis provided insights into the timing of crimes, revealing potential seasonal trends, and helping to forecast periods that may require increased vigilance.

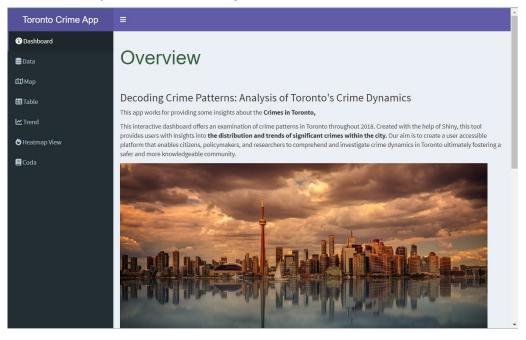
Heatmap Visualization: Premises Type and Crime Correlation

Employing plotly within our Shiny application, we created interactive heatmaps to examine the relationship between premises types and crime occurrences. This analysis shed light on which types of locations—be it apartments, commercial spaces, or outdoor areas—were more prone to specific crime categories, illustrating the nexus between environment and criminal behavior.

The amalgamation of these analytical approaches painted a comprehensive picture of crime in Toronto, serving as a basis for informed decision-making. From highlighting high-risk neighborhoods to identifying temporal crime patterns and premises vulnerable to offenses, our analysis aimed to arm stakeholders with the insights needed to deploy resources more effectively and enhance public safety strategies.

Coda

Hope you enjoy the tour of decoding crime patterns in Toronto with us. For more information, please check the <u>Shiny Dashboard</u>. Thank you.



Reference

Weisburd, D., Groff, E. R., & Yang, S. M. (2012). *The criminology of place: Street segments and our understanding of the crime problem.* Oxford University Press.