

# Temporal and Spatial Analysis of Hate Crimes in Toronto\*

Uncovering Neighborhood Disparities and Bias Trends Across the City

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This paper analyzes the temporal and spatial patterns of hate crimes in Toronto focusing on trends over time and neighborhood disparities. Toronto Open Data hate crime is used to explore variations in crime frequency and the prevalence of biases such as race, religion and sexual orientation. The results highlight distinct trends and reveal neighborhoods with higher concentrations of bias-motivated incidents; these findings offer insights for addressing hate crimes in the city.

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\*Code and data are available at: [https://github.com/YichengFu/hate\\_crimes.git](https://github.com/YichengFu/hate_crimes.git)

# 1 Introduction

Hate crimes are a significant incident reflecting deep-rooted prejudice and discrimination within communities. Hate-crime victimization against racially visible people is of growing concern (Chongatera 2013). These crimes not only impact individuals but also harm the society often leaving individuals feeling unsafe. Social violence omit both the daily violence suffered by certain social categories and its many impacts on the victims (Dalphond 2021). In Toronto, understanding the patterns and dynamics of hate crimes is crucial for addressing their causes and mitigating their effects. While racial and cultural diversity initiatives are central in hate crime policy, combating racially motivated hate crime is often obscured by matters considered more significant by police (Bryan 2019). This paper aims to fill that gap by providing an in-depth analysis of hate crimes in Toronto over time and across different neighborhoods.

This study focuses exploring two primary questions: how hate crimes in Toronto have evolved over time and whether certain neighborhoods experience a disproportionate concentration of bias-motivated incidents. Using Toronto open Data hate crime from 2018 to 2023 I analyzes temporal trends to identify peaks and patterns in the frequency of reported incidents. Additionally, the research examines whether specific biases (such as race, religion, or sexual orientation) are more prevalent in particular areas contributing to an understanding of neighborhood-level disparities. The results provide insight into both the temporal and spatial aspects of hate crimes, shedding light on how bias manifests in different contexts within the city.

The data section will introduced the detail of the data set used in this research. The variables as well as the cleaning process will be discussed in this section. The result section will focus on the findings and the visualization of the analysis. The discussion session talks about the limitation of this research and potential error caused by research design or the natural form of data. lastly, the conclusion part raps up the discovers and summarizes all the findings into a short paragraph.

## 2 Data

Some of our data is of penguins (**?@fig-bills**), from Horst, Hill, and Gorman (2020).

Talk more about it.

And also planes (**?@fig-planes**). (You can change the height and width, but don't worry about doing that until you have finished every other aspect of the paper - Quarto will try to make it look nice and the defaults usually work well once you have enough text.)

Table 1: Explanatory models of flight time based on wing width and wing length

|             | First model     |
|-------------|-----------------|
| (Intercept) | 1.12<br>(1.70)  |
| length      | 0.01<br>(0.01)  |
| width       | −0.01<br>(0.02) |
| Num.Obs.    | 19              |
| R2          | 0.320           |
| R2 Adj.     | 0.019           |
| Log.Lik.    | −18.128         |
| ELPD        | −21.6           |
| ELPD s.e.   | 2.1             |
| LOOIC       | 43.2            |
| LOOIC s.e.  | 4.3             |
| WAIC        | 42.7            |
| RMSE        | 0.60            |

**3 Results**

Our results are summarized in [Table 1](#).

**4 Discussion**

**4.1 First discussion point**

If my paper were 10 pages, then should be be at least 2.5 pages. The discussion is a chance to show off what you know and what you learnt from all this.

## **4.2 Second discussion point**

## **4.3 Third discussion point**

## **4.4 Weaknesses and next steps**

Weaknesses and next steps should also be included.

Examining how the model fits, and is affected  
by, the data

## References

Horst, Allison Marie, Alison Presmanes Hill, and Kristen B Gorman. 2020. *Palmerpenguins: Palmer Archipelago (Antarctica) Penguin Data*. <https://doi.org/10.5281/zenodo.3960218>.