

# Analysing Road Safety

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Erdős Institute project

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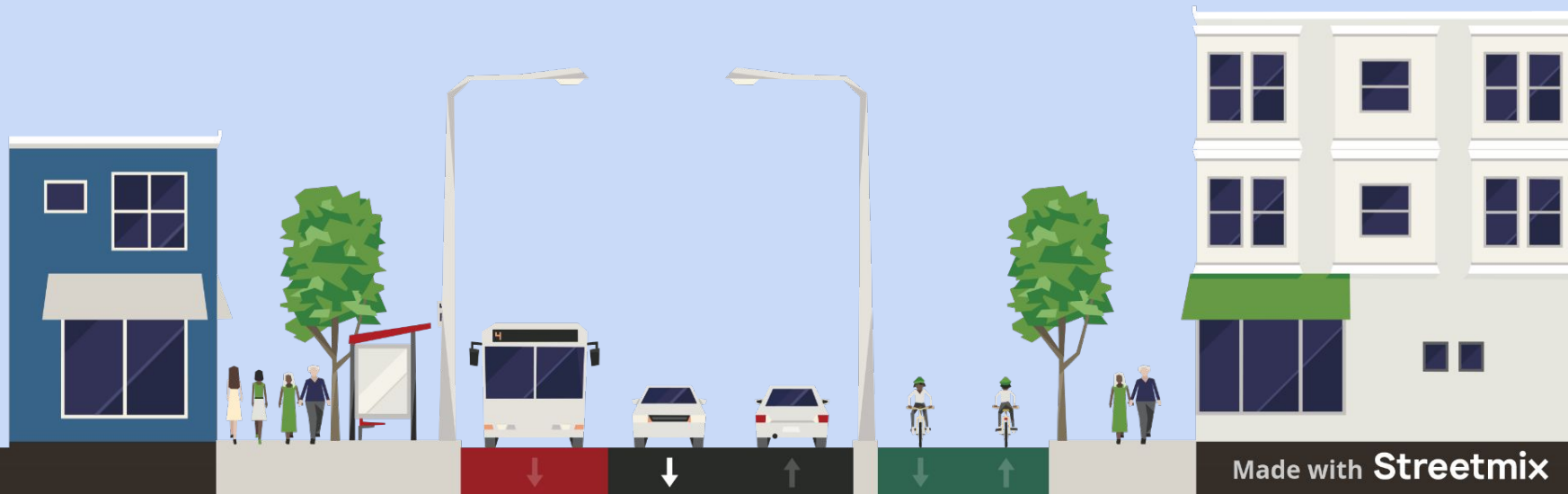
# Overview



Increasingly, cities across North America are adopting postures that injuries and deaths on roadways are not inevitable, but *preventable*, and that safety is a property of a street's design, and less a consequence of enforcement. This is commonly referred to as “Vision Zero”.

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# Goal: Which changes make streets safer?



# Goal: Predict collision rate from physical design

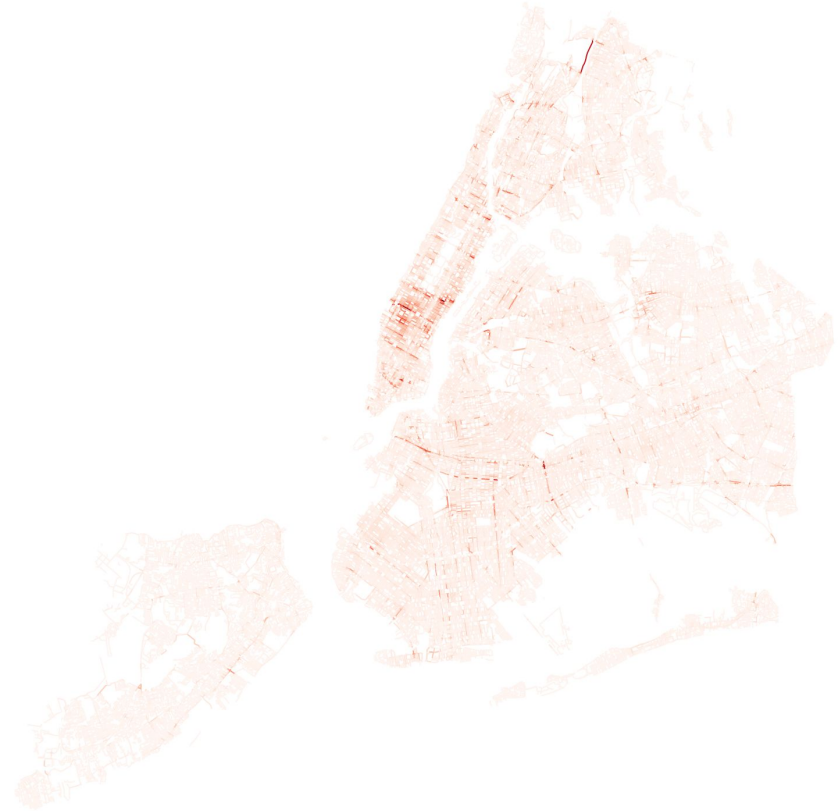


# New York City

This city has both a **Vision Zero** policy and many publicly available data.

## Stakeholders:

- Residents
- City Planners
- Neighbourhood advocacy groups



(Higher collision rates in darker red)

# Datasets

NYC Open Data

**NYC** OpenData

- Street width
  - Weekly collision rate
  - Presence of bike lanes
  - Number of trees planted
  - Traffic volume
  - Presence of speed humps
  - Presence of parking meters
  - Speed limit
  - ... and more
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# Data Processing

Selecting streets whose traffic  
volume is measured

Street segments with traffic  
volume data shown in blue.



# Data Processing

Temporal and spatial  
transformations

- Streets which underwent a recent change (e.g. installing speed humps) were split into multiple entries.
  - Geographic data was assigned to the street it lay on.
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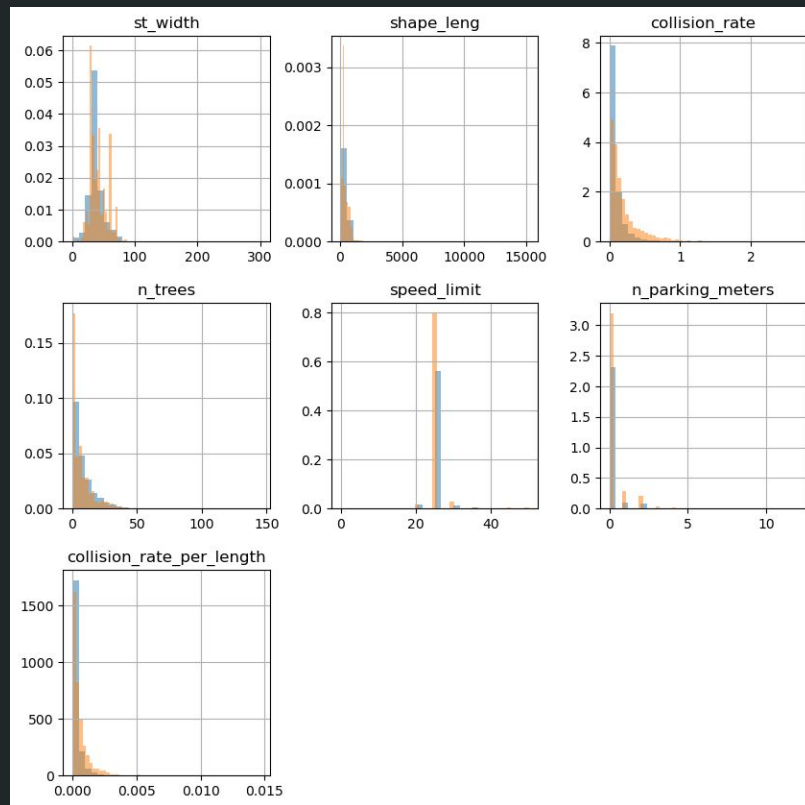
# Initial Observations

## Observations on collisions:

- Relatively rare
  - Happen in city centers
  - Happen along certain corridors
  - Concentrated on a few streets
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# Initial Observations

Streets with volume data are similar to ones without

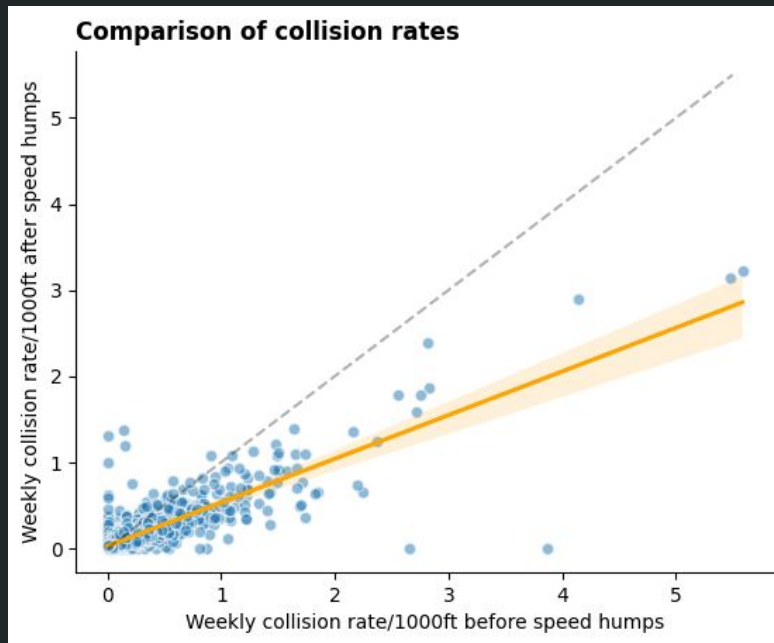


Distributions of volume vs no volume data

# Initial Observations

Speed hump analysis

For roads with recent speed hump installations, we observed a trend consistent with other research:



# Models

- **Baseline constant model**
  - ***k*-nearest neighbours**
  - **Linear regression**
  - **Random forest with XGBoost**
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# Model Performance

Best model:  
K-Nearest Neighbours

Model	Mean Square Error ( $\times 10^{-7}$ )
Baseline	4.85
Linear regression	3.83
Random forest	3.56
K-Nearest Neighbours	3.48

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# Feature Importance

Using random forest to determine feature importance

## Top 5 important features:

- Traffic Volume
- Street Width
- Length of road segment
- Number of Trees
- Type of street (ave or st)

## Less impactful features:

- Speed limit
  - Has a bike lane
  - Parking meters
  - Speed humps
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# Final thoughts

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