

Introduction

- In their most recent report, the World Health Organization estimated that global deaths due to road traffic was over **1.3** million people per year.
- Countries have reduced this burden by implementing road, vehicle and user safety standards strategies.
- Many traffic collision **features can influence an injury outcome** people, vehicle, environment, infrastructure, fuel price and climate change elements are relevant.
- Supervised machine learning is suitable for relating diverse predictors to traffic collision injuries; previous work have used Decision Tree (**DT**) and k-Nearest Neighbour (**kNN**) methods.

Research Questions

- 1. How can Ottawa collision injuries be described top locations, severity distribution per mode of transportation, occurrence correlation, severity correlation, and traffic control implications?
- 2. Can we build a model to **predict the likelihood of injury** using Decision Tree and k-Nearest Neighbour methods?

Methods

- **Data Source**: <u>https://open.ottawa.ca/</u> accessed in September 2023.
- Preprocessing
 - Missing integer values were changed to zeros.
 - Latitude/Longitude values were truncated to 4 decimal points.
 - A binary injury variable was created; 0 = no injury, 1 = at least one injury.
- Ordinal features were converted to integer scales.
- Nominal features were converted to one-hot encoding.
- Descriptive Analysis
 - Correlation heat maps used **Pearson correlation coefficients**.
- Modeling
- Packages (Python): SciKit-Learn, imblearn.over_sampling
- 80-20 Train-Test split with nested 70-30 Train-Validate split of outer Train set.
- kNN
 - Hyperparameters: nearest neighbours = 6
 - Class imbalance was addressed using Random OverSampling and SMOTE.
- DT
 - Hyperparameters: entropy criterion, max depth = 6
 - Class imbalance was addressed using Random OverSampling.
 - Random Forest (n_estimators = 100)
- Model Evaluation
- DT was evaluated against a variety of techniques using PyCaret [Python].

Acknowledgements

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Figure 1. Top 10 Locations in Ottawa, Ontario with the most collisions and the most collisions involving at least one injury. Top: geographic maps marked by the locations with the most collisions (left) and most collisions with injuries (right). Bottom: Frequency tables of locations with the most collisions (left) and the most collisions with injuries (**right**).

	Collisions			Collisions with an Injury					
Rank	Location	n	Rank	Location	n				
1	Hunt Club Rd @ Riverside Dr	232	1	Meadowlands Dr @ Merivale Rd	33				
2	St. Joseph Blvd @ Jeanne D'arc Blvd	218	2	Hunt Club Rd @ Riverside Dr	32				
3	Innes Rd @ Tenth Line Rd	168	3	Riverside Dr @ Tremblay Rd/Hwy417 Ic117 Ramp52	32				
4	West Hunt Club Rd @ Woodroffe Ave	157	4	Innes Rd @ Tenth Line Rd	32				
5	Bank St @ Walkley Rd	157	5	Prince Of Wales Dr @ West Hunt Club Rd	29				
6	Prince Of Wales Dr @ West Hunt Club Rd	155	6	Hazeldean Rd @ Terry Fox Dr	28				
7	Baseline Rd @ Woodroffe Ave	155	7	Baseline Rd @ Woodroffe Ave	28				
8	Greenbank Rd @ Strandherd Dr	147	8	Hunt Club Rd @ Bridle Path Dr/Daze St	28				
9	St. Joseph Blvd/Old Montreal Rd @ Trim Rd	147	9	Albion Rd @ Mitch Owens Rd	27				
10	Merivale Rd @ West Hunt Club Rd	146	10	Innes Rd @ Jeanne D'arc Blvd/Mer Bleue Rd	27				

Ottawa Traffic Collisions – Describing and Predicting Injuries with Supervised Learning

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injury	-0.08	0.99	0	-0.06	-0.04	-0.06	-0.08	0.04	0.29	0.24	0.14	0	0.03	1
long_4	-0.07	0.02	-0.08	-0.01	-0.01	-0.01	-0.07	0.07	0.01	0.01	0	0.4	jt.	0.03
lat_4	-0.02	0	-0.02	-0.01	-0.01	-0.02	-0.02	0.02	0.01	0.01	0		0.4	0
num_of_motorcycles	0	0.14	0.04	-0.03	-0.04	-0.02	0.01	-0.03	-0.01	-0.01	1	o	0	0.14
num_of_bicycles	-0.05	0.23	-0.05	-0.06	-0.05	-0.04	-0.04	0.04	-0.02	1	-0.01	0.01	0.01	0.24
num_of_pedestrians	-0.06	0.3	0.22	-0.03	-0.02	0.01	-0.06	-0.2	1	-0.02	-0.01	0.01	0.01	0.29
num_of_vehicle	-0.27	0.03	-0.64	-0.1	-0.08	-0.29	-0.27	1	-0.2	0.04	-0.03	0.02	0.07	0.04
traffic_control	0.98	-0.08	0.3	0.05	0.04	0.18	1	-0.27	-0.06	-0.04	0.01	-0.02	-0.07	-0.08
light	0.18	-0.06	0.26	0.11	0.09	1	0.18	-0.29	0.01	-0.04	-0.02	-0.02	-0.01	-0.06
environment_condition	0.04	-0.04	0.07	0.56	1	0.09	0.04	-0.08	-0.02	-0.05	-0.04	-0.01	-0.01	-0.04
road_surface_condition	0.05	-0.06	0.07	1	0.56	0.11	0.05	-0.1	-0.03	-0.06	-0.03	-0.01	-0.01	-0.06
initial_impact_type	0.3	0	1	0.07	0.07	0.26	0.3	-0.64	0.22	-0.05	0.04	-0.02	-0.08	0
classification_of_accident	-0.08	1	0	-0.06	-0.04	-0.06	-0.08	0.03	0.3	0.23	0.14	0	0.02	0.99
location_type	t	-0.08	0.3	0.05	0.04	0.18	0.98	-0.27	-0.06	-0.05	0	-0.02	-0.07	-0.08
location type accident act type condition light control vehicle strans picycles lat a long a minny traffic control restricted by the property of the property														

earson Correlation 0.5 -0.5

Figure 2. Correlation matrix of all predictor variables and injury occurrence.

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Top 10 Locations by Injury Count



Rank	Location					
Precision	0.88					
Recall	0.24					
F1 Score	0.37					
Accuracy	0.85					

Results

1. The "Hunt Club Bridge" is one of the most collision/injury prone locations (Figure 1). 2.Six locations had 3 major injuries. Three locations had 2 fatal injuries. Both outcomes involved 8th Line Rd @ Parkway Rd.

3. Mode of transportation features (motorcycles, pedestrians, bicycles) correlated with an injury occurring (Figure 2).

4.**Environment** features did not correlate with injury severity.

5.Most collisions involved only Vehicles (95.28%) and of these most were without injury (80%). Pedestrian collisions almost always involved injury (> 99%). 3.33% of **Motorcycle** injuries were *fatal* while 2.43% of **Pedestrian** injuries were *fatal*. 6.Combined, collisions usually occurred when there was a traffic control (signals, stop signs, etc.). Separated, the most common type of traffic control was no control.

7.Compared to the best benchmark model (Ridge Classifier), **Decision Tree** model (Figure 3) had the highest accuracy (0.85 > 0.84) and **precision** (0.88 > 0.69) but not **recall** (0.24 < 0.26).

8.kNN performed worse than chance. Random Forest performed worse than the final decision tree model.

Implications

- First responders can better anticipate the degree of injury with this model.
- High collision/injury locations may deserve targeted **interventions**.
- Environmental features may not correlate to traffic collision injuries when analyzed categorically.

Future Work

- **Time** analysis does time of day, day of week, seasonality, COVID affect injury incidence?
- Incorporate **traffic volume** data for more comprehensive model construction.
- Intervention analysis did traffic **camera installation** result in fewer collisions and injuries?

Conclusion

- Multiple Ottawa locations have high collision frequency, with the "Hunt Club Bridge" being most prevalent.
- 8th Line Rd @ Parkway Rd. is a dangerous traffic location.
- Mode of transport affects the occurrence and type of injury.
- **Decision Tree** is a useful method for predicting traffic collision injuries with this dataset.

References

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Figure 3. Decision Tree model test measures (left) and confusion matrix (right).