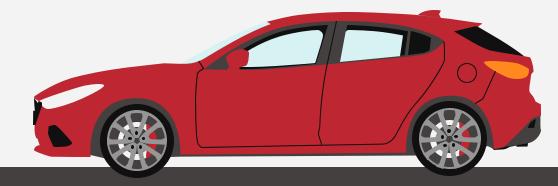
# Predicting Traffic Accident Severity using KNN

By: Jasmine Ly



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## Introduction

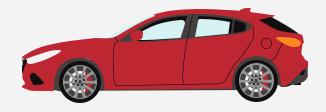
**Objective:** Develop a KNN model to predict traffic accident severity using key features (distance, weather conditions, and visibility)

Dataset: US Accidents (2016–2023) from Kaggle.

• Weather-related and numerical features

#### Why?

- Identify patterns in severity across accidents in the US
- Better understanding will allow for creating better prevention and response



# Methodology

**Target Variable:** Severity

#### **Features Used**

- Distance (mi)
- Wind Speed (mph)
- Precipitation (in)

- Visibility (mi)
- Temperature (F)
- Humidity (%)

#### **Process**

- Quantitative, predicts accident severity
- Used 5% of dataset for the model to prevent errors
- Split the data 80/20 training/testing
- **for** loop tested different k values and created an elbow graph of the results for analysis
- Highest accuracy was 0.67 for k=59

# Challenges

#### **Extremely Large Data Set**

- Original dataset contains 7.7 million records and 3.06 GB in size
- This lead to extremely long processing times and memory errors

#### **Data Quality Issues**

- Columns with mixed data types created preprocessing challenges
- Missing/incomplete data needed cleaning, which could potentially lead to a loss of valuable information.

#### **Feature Selection**

- Due to scale of project, we had to narrow down to only 6 factors
- Broadly applicable across regions

## **Lessons Learned**

#### **Limitations of KNN**

 While KNN is a straightforward model that is easy to implement, it struggles to compute large multidimensional datasets due to its simplicity

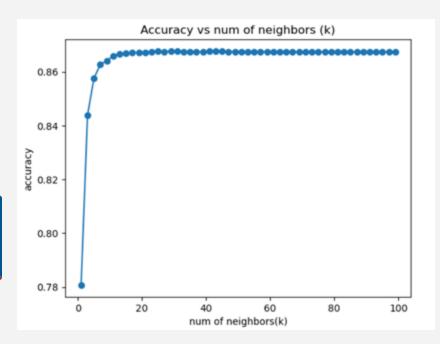
#### Importance of Data Quality

 Data cleaning and preprocessing are critical steps of building a reliable model.

# Value of Hyperparameter Tuning

 Experimenting with different values of K and plotting the results showed how tuning hyperparameters can optimize model performance.

# **Results**



Trained on 7.7 million records

#### **Performance Measures**

Accuracy: 86.7%Precision: 79%Recall: 86.7%

#### **Largest Mutual Coefficients**

Distance: 0.123Visibility: 0.225Wind Speed: 0.018

# Conclusion

• **Project Objective**: Developed a K-Nearest Neighbors (KNN) model to predict the severity of traffic accidents based on critical features like time of distance, weather conditions, and visibility.

#### Key Processes:

- Data Preprocessing: Cleaned and prepared the dataset, including handling missing values and encoding categorical features.
- Exploratory Data Analysis (EDA): Identified key patterns and relationships in the data.
- Model Tuning: Optimized hyperparameters to improve accuracy and performance.
- Evaluation: Assessed the model using relevant metrics to validate predictions.

#### Results & Impact:

- Identified factors correlating with high accident severity.
- Demonstrated the potential to inform road safety policies and improve emergency response.
- Showcased how predictive analytics can enhance public safety.





# Thank you

