Machine Learning: Car Accident Predictions

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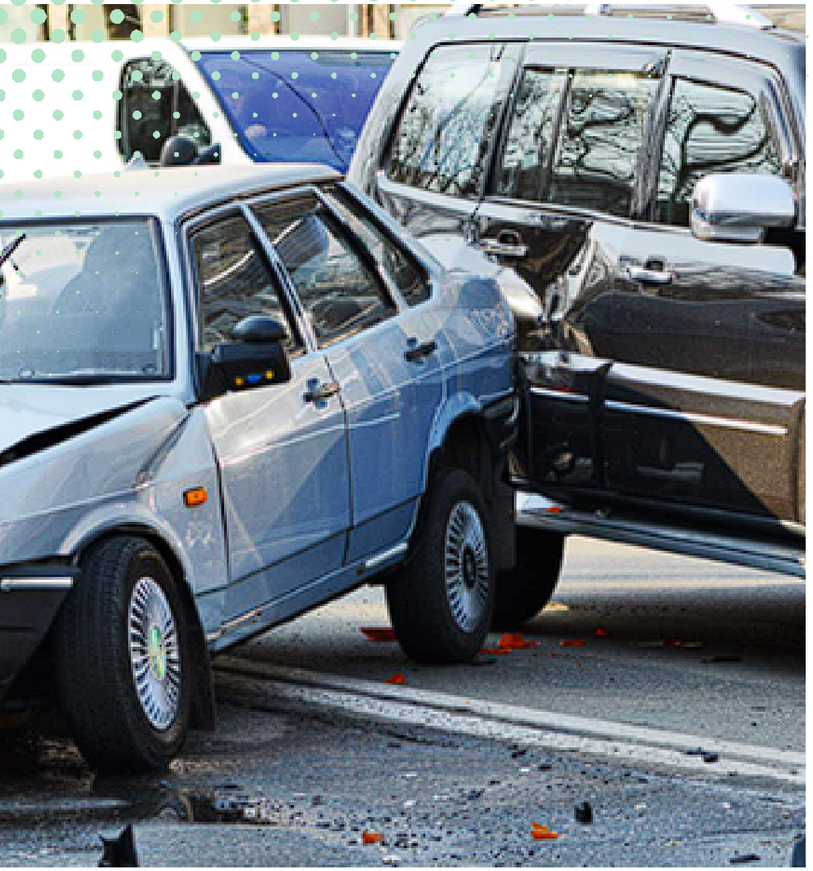
Agenda

- Project overview
- Description of data and purpose
- Data preprocessing
- Machine learning models
- Visualizations
- Lessons learned



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Project Overview

In this project we tried to classify the severity of road accidents based off of different factors (sex, weather, time of day,etc). The data was collected from vehicle collisions in Ethiopia ranging from 2017-2020.

Road accidents are one of the major causes of unnatural deaths around the world. We used Machine Learning with the objecting of measuring how influential certain factors can be in accidents taking place in the hopes of reducing the amount and severity in the future.

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Discription of Data

Dataset:

- 32 columns and 12316 rows.
- Accident Severity broken into 3 categories: slight injury, serious injury fatal injury.
- Other factors considered:
  Road surface, Time of Day,
  Weather, Sex of Diver, Age of Driver .

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Target variable

Accident severity

- Minor
- Severe
- Fatal

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Data Preprocessing steps

- Dataset from Kaggle
- Import libraries
- Identifying missing values
- Convert Time to datetime and extract hour
- Encoding Categorical data
- Concat 2 Data Frames
- Split dataset
- Standard scaling
- Train models
- Prediction

Random Forrest Model

- Used to handle large datasets with many features
- Identify important features to help identify which variables are most predictive of outcome
- Ability to handle both numerical and categorical data types
- Supervised Model

Random Forrest Model

85.5% Accuracy

Confusion Matrix

	Predicted 0	Predicted 1	Predicted 2
Actual 0	2629	1	0
Actual 1	407	5	0
Actual 2	37	0	0

Accuracy Score : 0.8554725560246833
Classification Report

	precision	recall	f1-score	support
0	0.86	1.00	0.92	2630
1	0.83	0.01	0.02	412
2	0.00	0.00	0.00	37
accuracy			0.86	3079
macro avg	0.56	0.34	0.32	3079
weighted avg	0.84	0.86	0.79	3079

Random Forrest Model

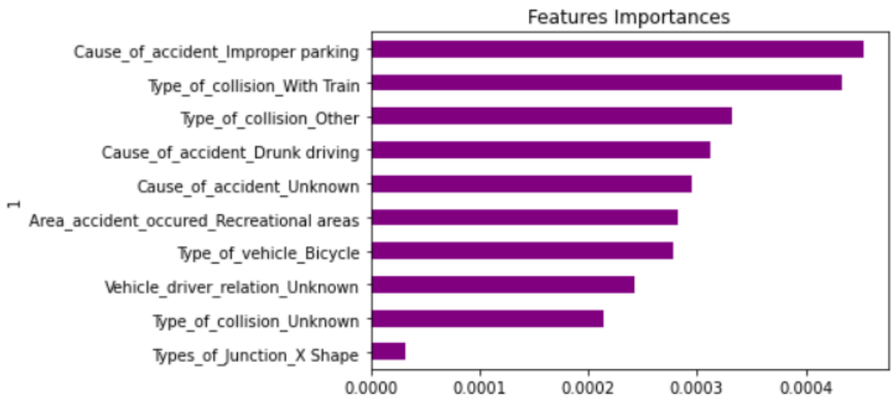
In [19]:
```
# Get the feature importance array
importances = rf_model.feature_importances_
# List the top 10 most important features
importances_sorted = sorted(zip(rf_model.feature_importances_, X.columns), reverse=True)
importances_sorted[:10]
```

Out[19]:
```
[(0.09165925304127531, 'Hour_of_Day'),
 (0.044992706371324256, 'Number_of_vehicles_involved'),
 (0.03416120280696221, 'Number_of_casualties'),
 (0.02076171882617486, 'Types_of_Junction_Y Shape'),
 (0.02026492335463812, 'Age_band_of_driver_31-50'),
 (0.01991774619550848, 'Educational_level_Junior high school'),
 (0.01943745759610502 5, 'Area_accident_occured_Other'),
 (0.01917118093411732, 'Types_of_Junction_No junction'),
 (0.01909525280974353, 'Service_year_of_vehicle_Unknown'),
 (0.01885960108745316 6, 'Sex_of_casualty_Male')]
```

Random Forrest Model

In [33]:
```
# Visualize the features by importance
importances_df = pd.DataFrame(sorted(zip(rf_model.feature_importances_, X.columns), reverse=True))
importances_df.set_index(importances_df[1], inplace=True)
importances_df.drop(columns=1, inplace=True)
importances_df.rename(columns={0: 'Feature Importances'}, inplace=True)
importances_sorted = importances_df.sort_values(by='Feature Importances')
importances_sorted=importances_sorted[:10]
importances_sorted.plot(kind='barh', color='purple', title= 'Features Importances', legend=False)
```

Out[33]: <AxesSubplot:title={'center':'Features Importances'}, ylabel='1'>



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K-Nearest Neighbors Model

- Supervised learning
- It is a simple and easy-to-understand
- Used in both regression and classification predictive problems

K-Nearest Neighbors Model

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- Trained to predict the severity of the accident based on certain features such as weather conditions, road type, time and etch.
- K=2 neighbors accuracy 86%

```
# Print classification report
print(classification_report(y_pred,y_test))
```

	precision	recall	f1-score	support
0	0.98	0.87	0.92	2938
1	0.17	0.52	0.25	132
2	0.11	0.44	0.17	9
accuracy			0.86	3079
macro avg	0.42	0.61	0.45	3079
eighted avg	0.94	0.86	0.89	3079

Visualizations

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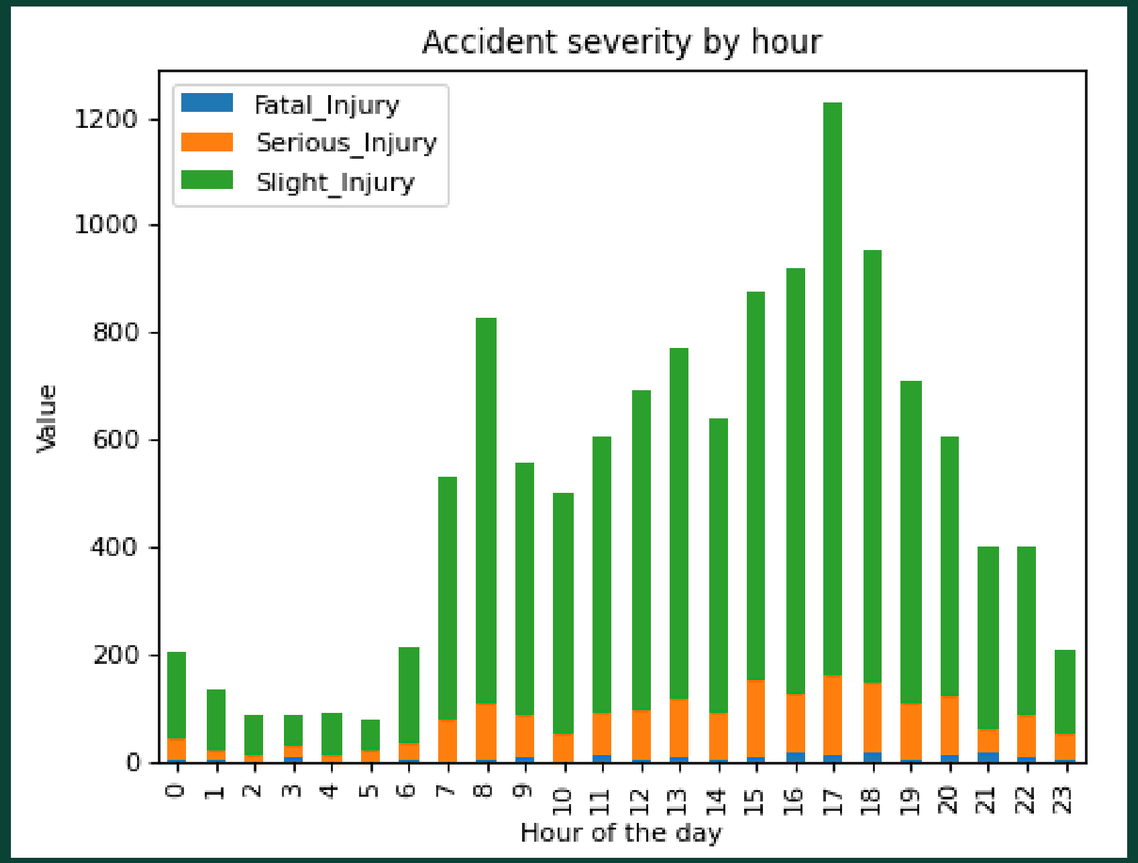
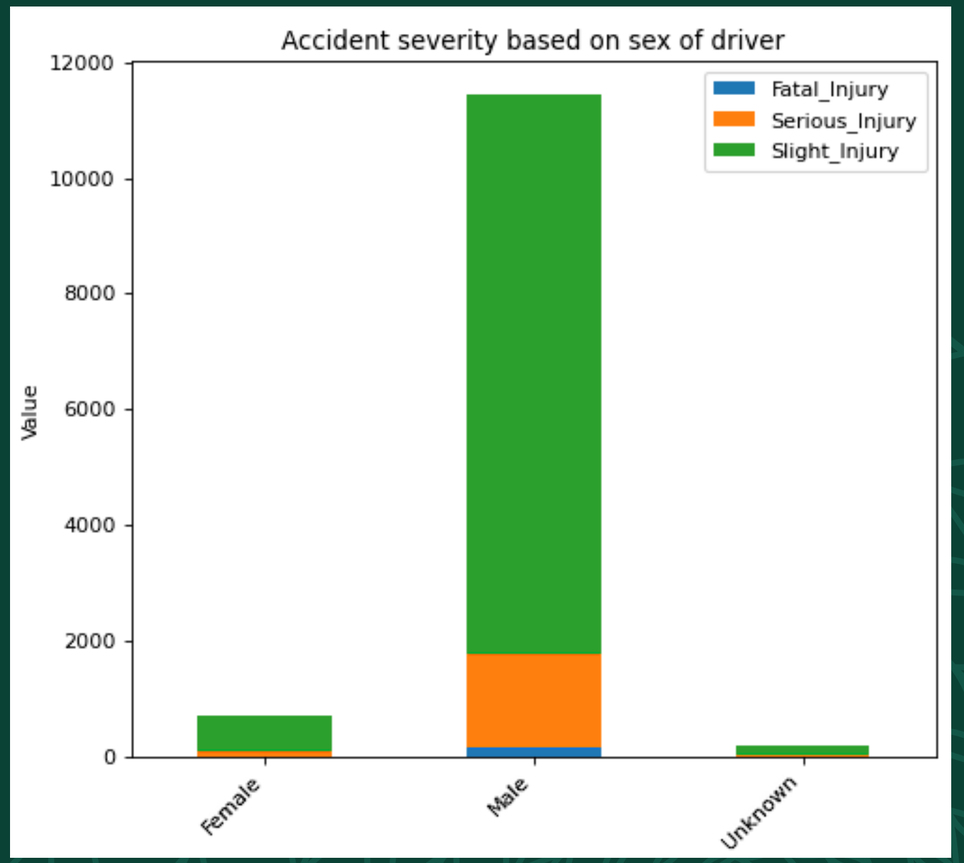


5PM is statistically the worst time to drive bases on this dataset



There is an extreme difference between the amount of accidents females and males are involved in. Based off this dataset

Visualizations

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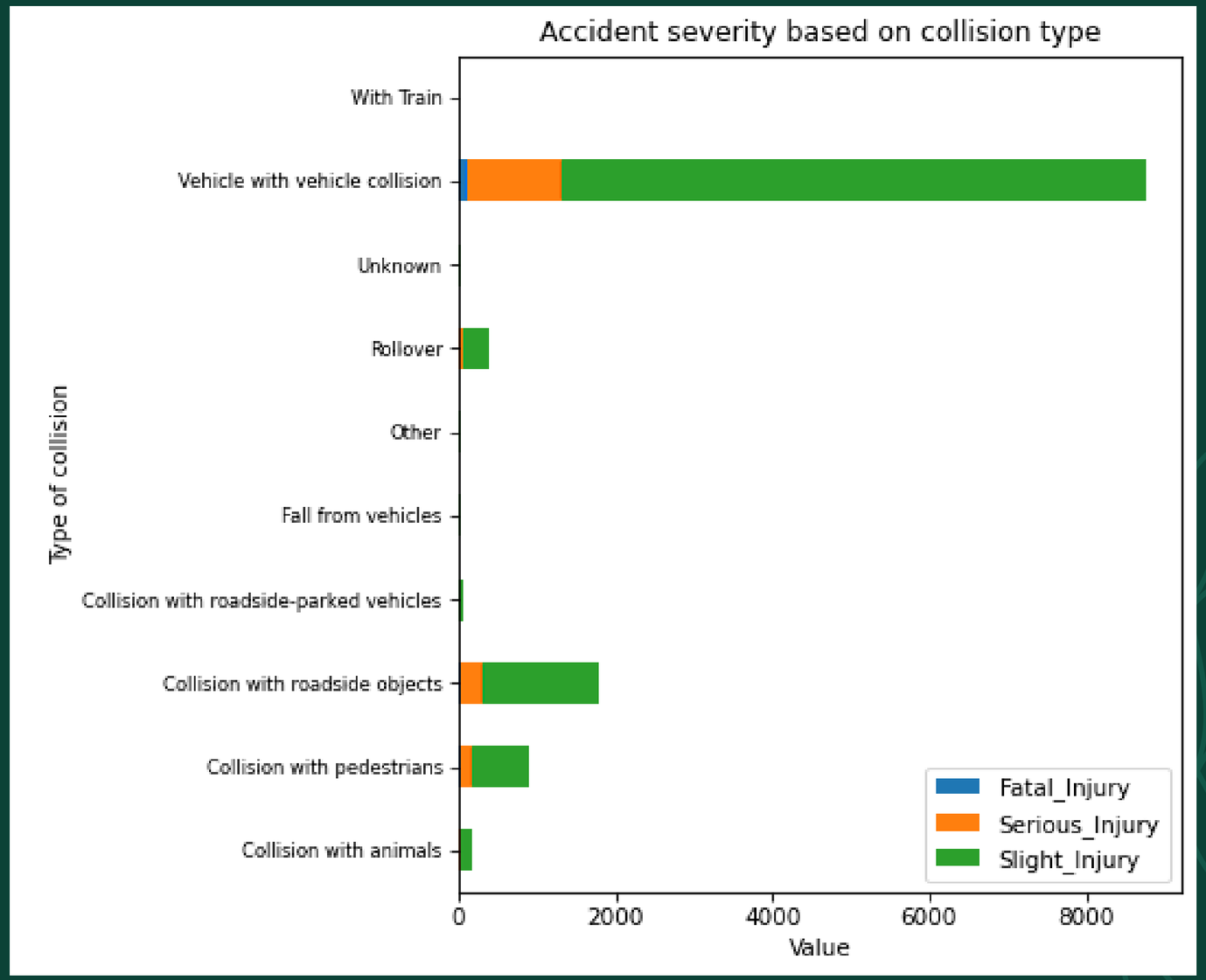

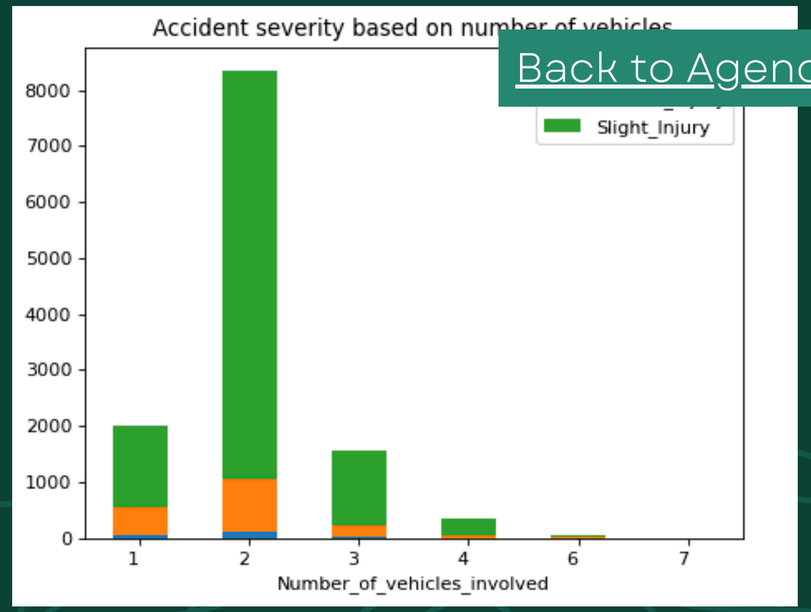

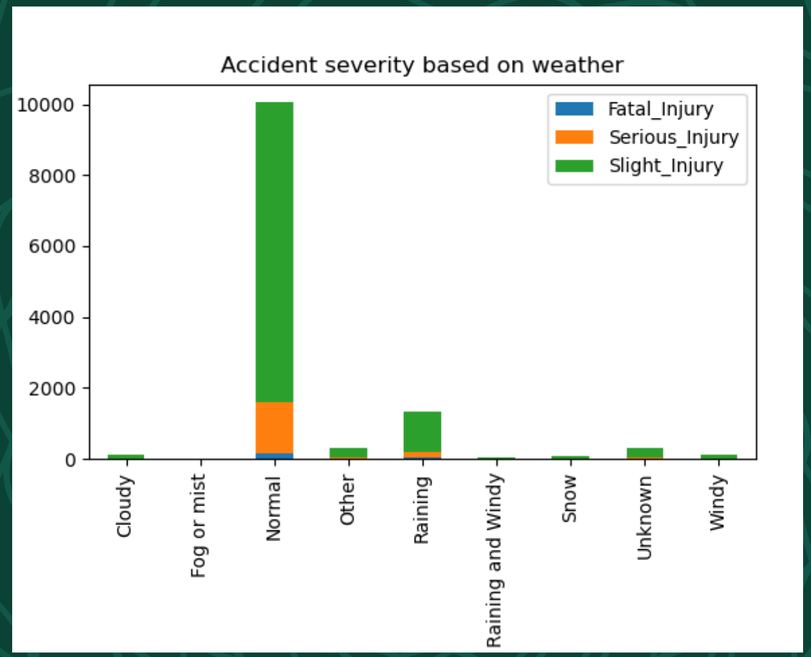

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