





# 1. Business Understanding

## A complex, hand-drawn mind map centered on 'WEB' and 'CLOUD', illustrating the interconnectedness of various digital technologies and concepts. The map branches out into categories like 'BIG DATA', 'SOCIAL', 'PEOPLE', 'DEVICES', 'APPS', 'PLATFORM', 'DEVELOPERS', 'SECURITY', 'IT', 'SQL', 'STORAGE', 'AZURE', 'CLOUD', and 'WEB'. It includes numerous icons representing data, communication, hardware, and software.

nce companies by getting useful insights  
ts. The Machine Learning model and its  
ful decisions for reducing the number of  
with various supervised machine learning

- accidents and injuries for the city. The most accurate algorithms i.e.
- **Algorithm A. Logistic regression**
  - **Algorithm B. The K-Nearest Neighbours**
  - **Algorithm C. Decision Tree**
  - **Algorithm D. Random Forest**
- And finally, the accuracy of each algorithm**

**And finally, the accuracy of each algorithm will be plotted to check which algorithm performs better.**



## 2. Data Understanding



## 2. DATA UNDERSTANDING



The data used for this project was collected by the SDOT traffic management Division and Seattle Traffic Records Group from 2004 to present. It was downloaded from the link shared in the IBM Applied Data Science Capstone course. The data consists of 38 independent variables and 194,673 rows. The dependent variable, “SEVERITYCODE”, contains numbers that correspond to different levels of severity caused by an accident from 1 to 2. Severity codes are as follows:

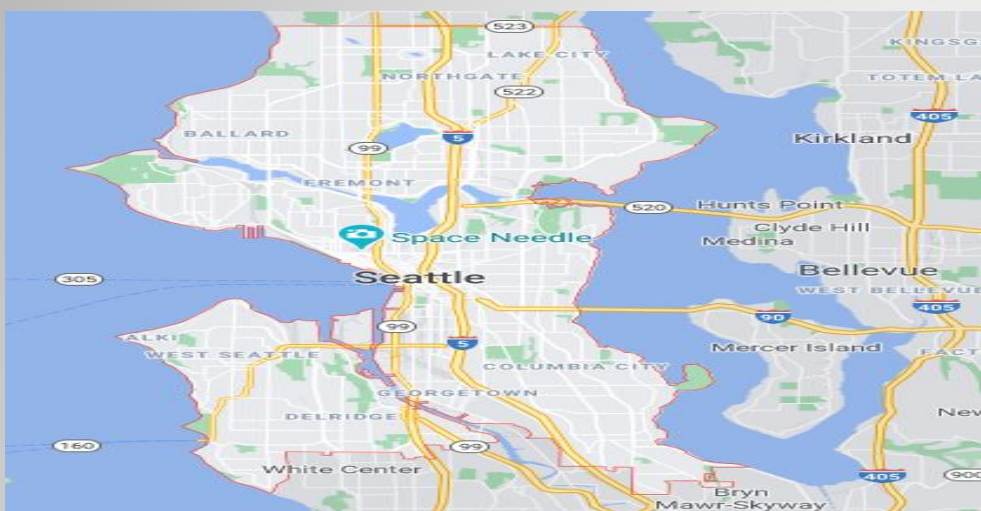
1: Property Damage Only Collision

2: Injury Collision

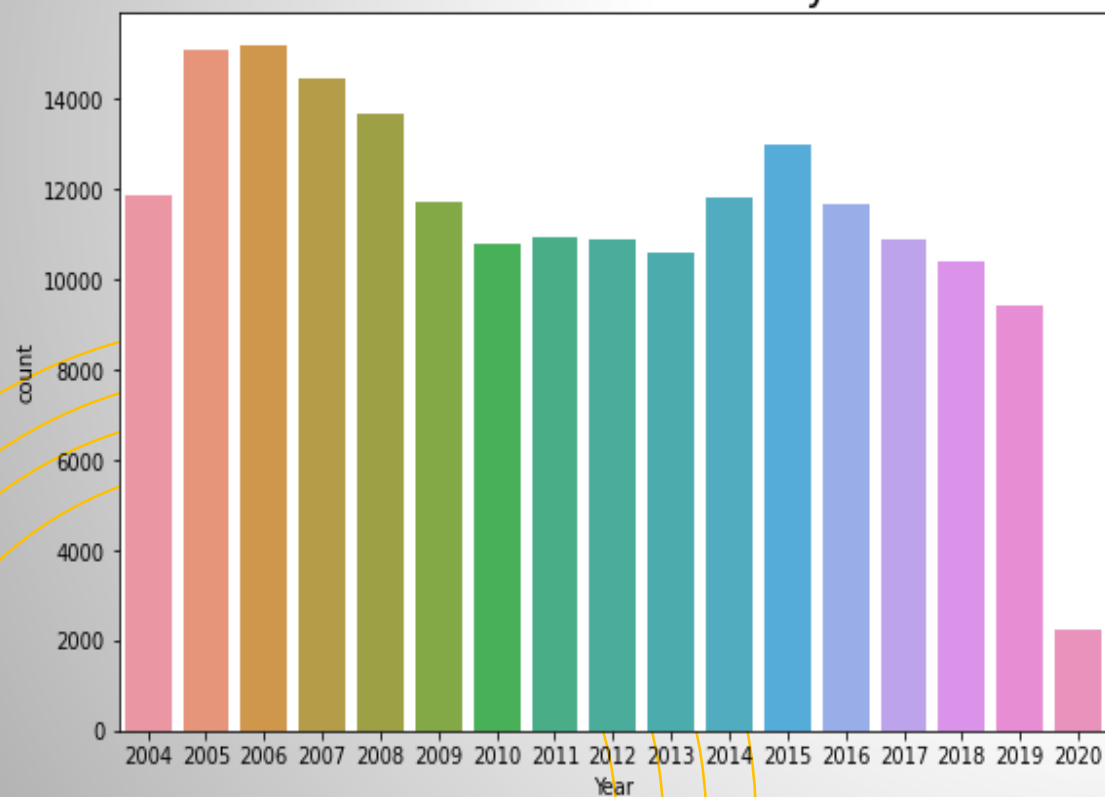
Furthermore, as there are null values in some records, hence the data has been pre-processed.

**The link for the Metadata is mentioned below:**

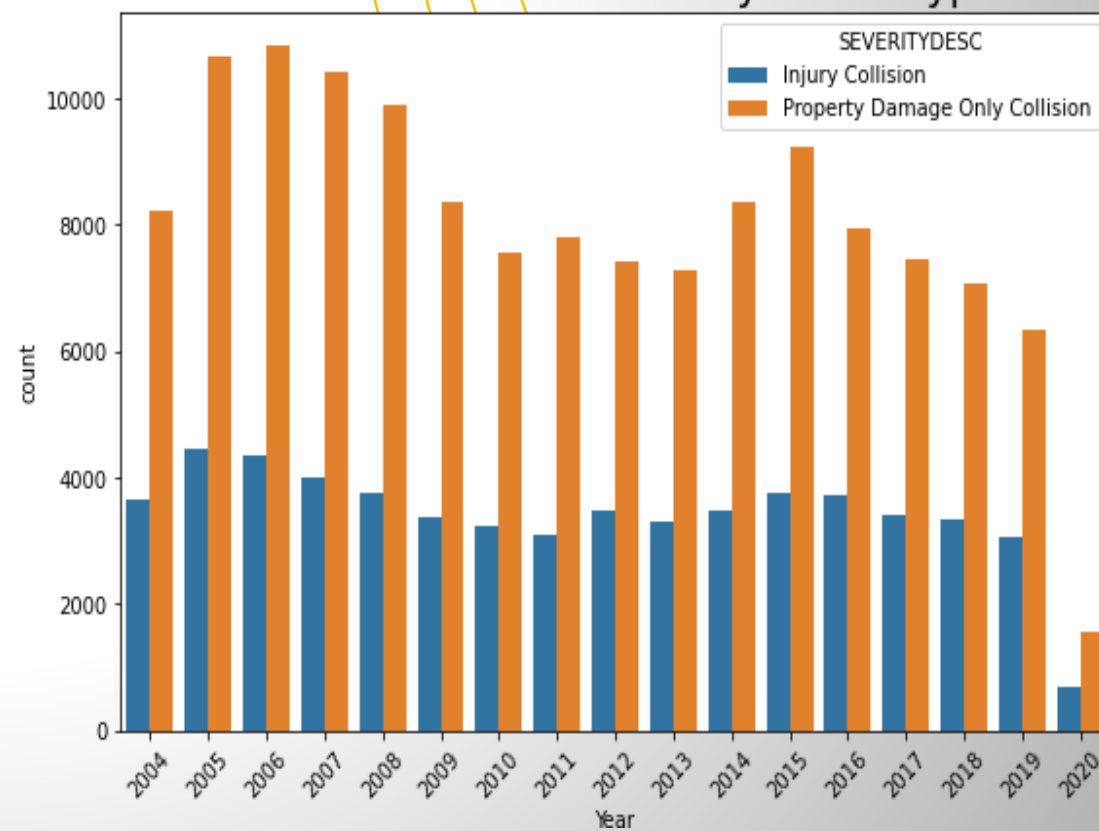
[https://github.com/Engineer00/Coursera\\_Capstone/blob/master/Scripts/Metadata.pdf](https://github.com/Engineer00/Coursera_Capstone/blob/master/Scripts/Metadata.pdf)



## Car accidents in Seattle by Year



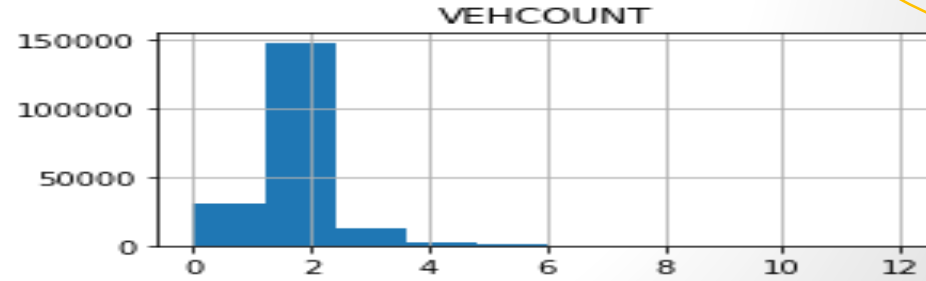
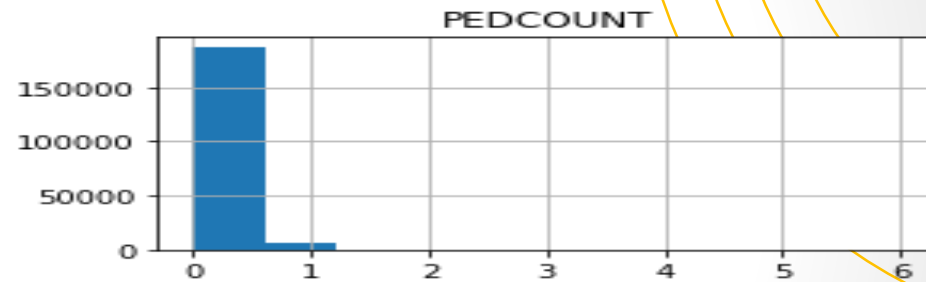
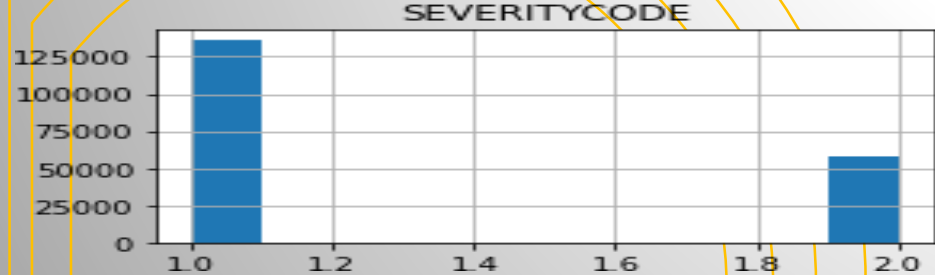
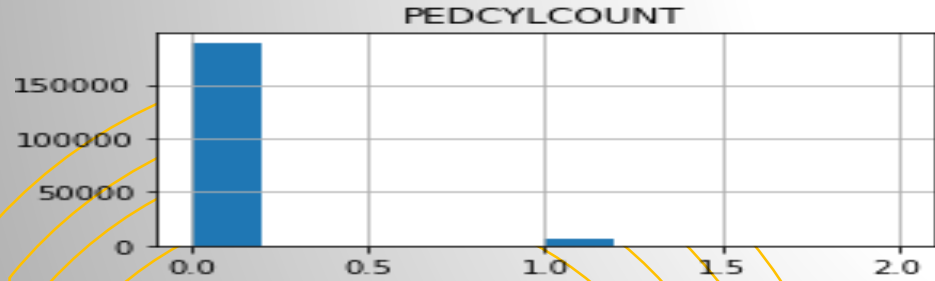
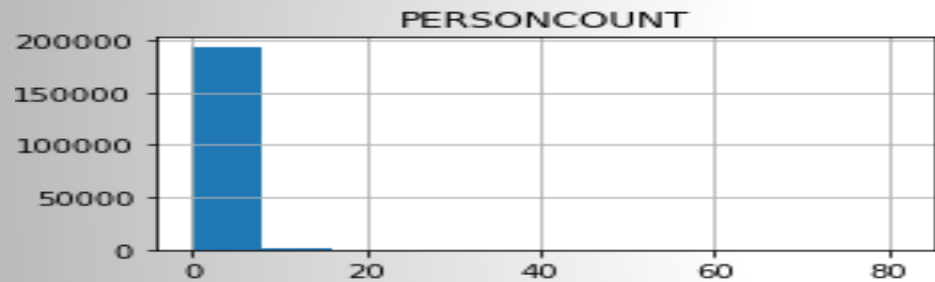
## Car accidents in Seattle by Year & type



# NUMERIC FEATURES VISUALIZATION

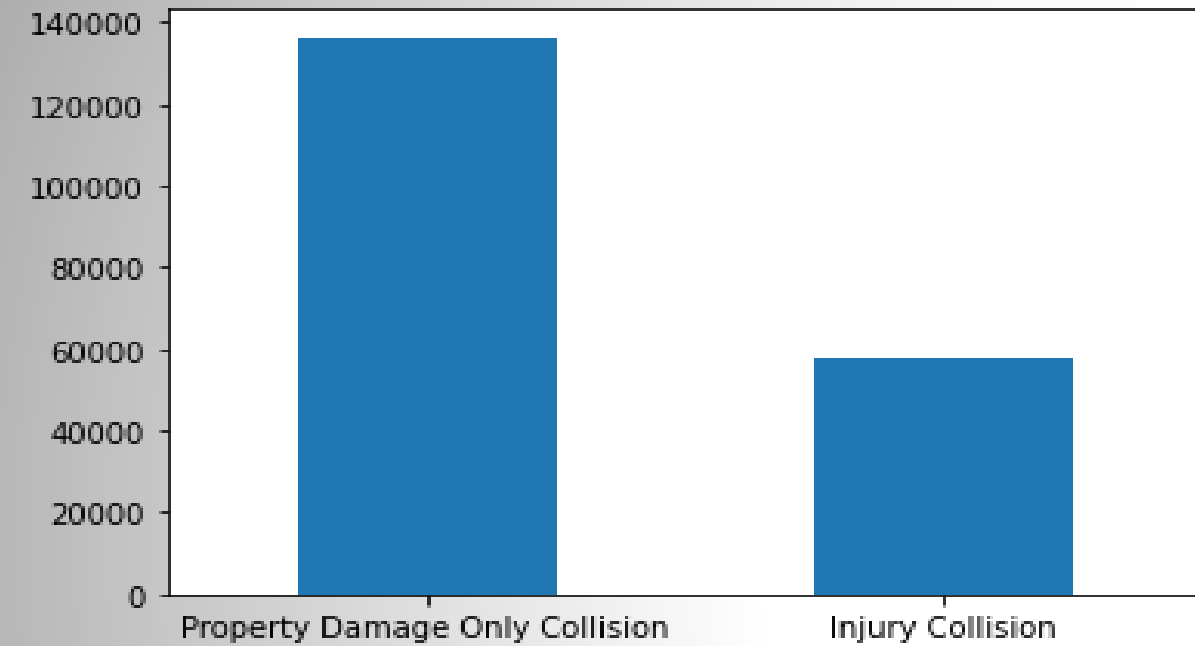


## Numeric features distribution

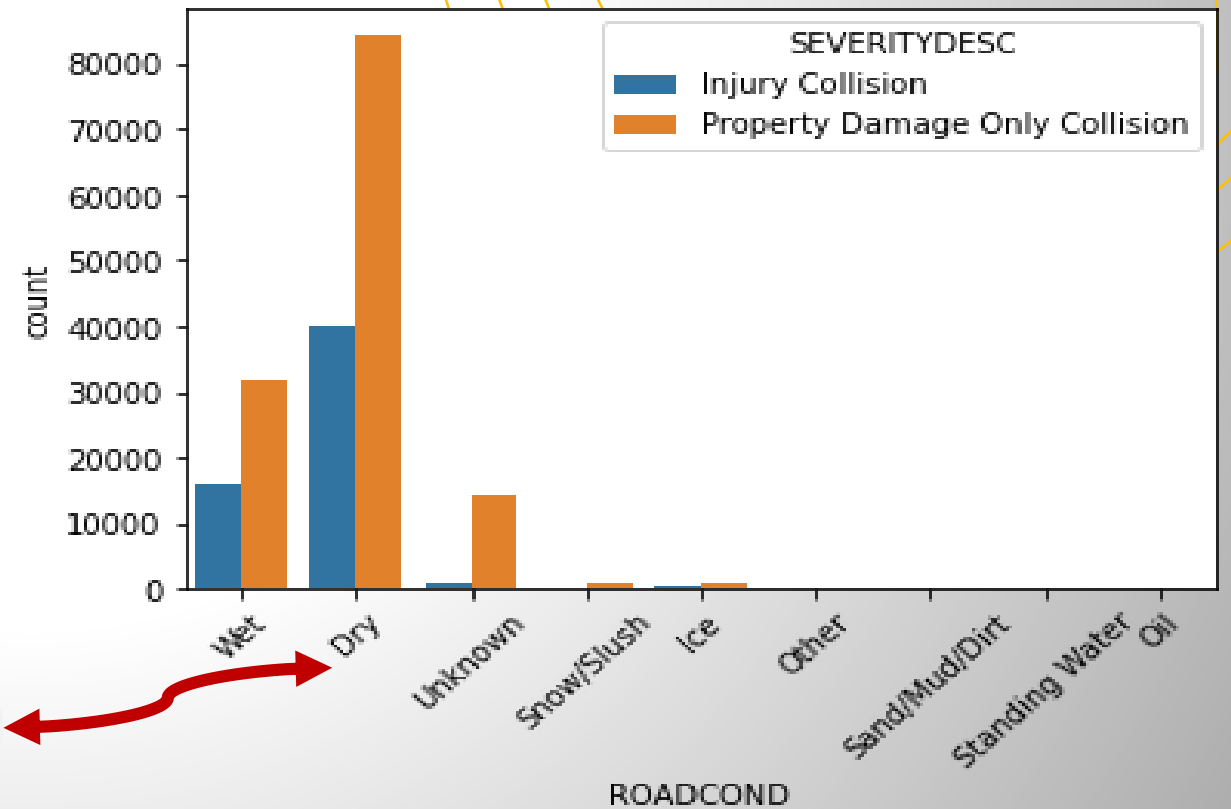




## CATEGORICAL FEATURES VISUALIZATION

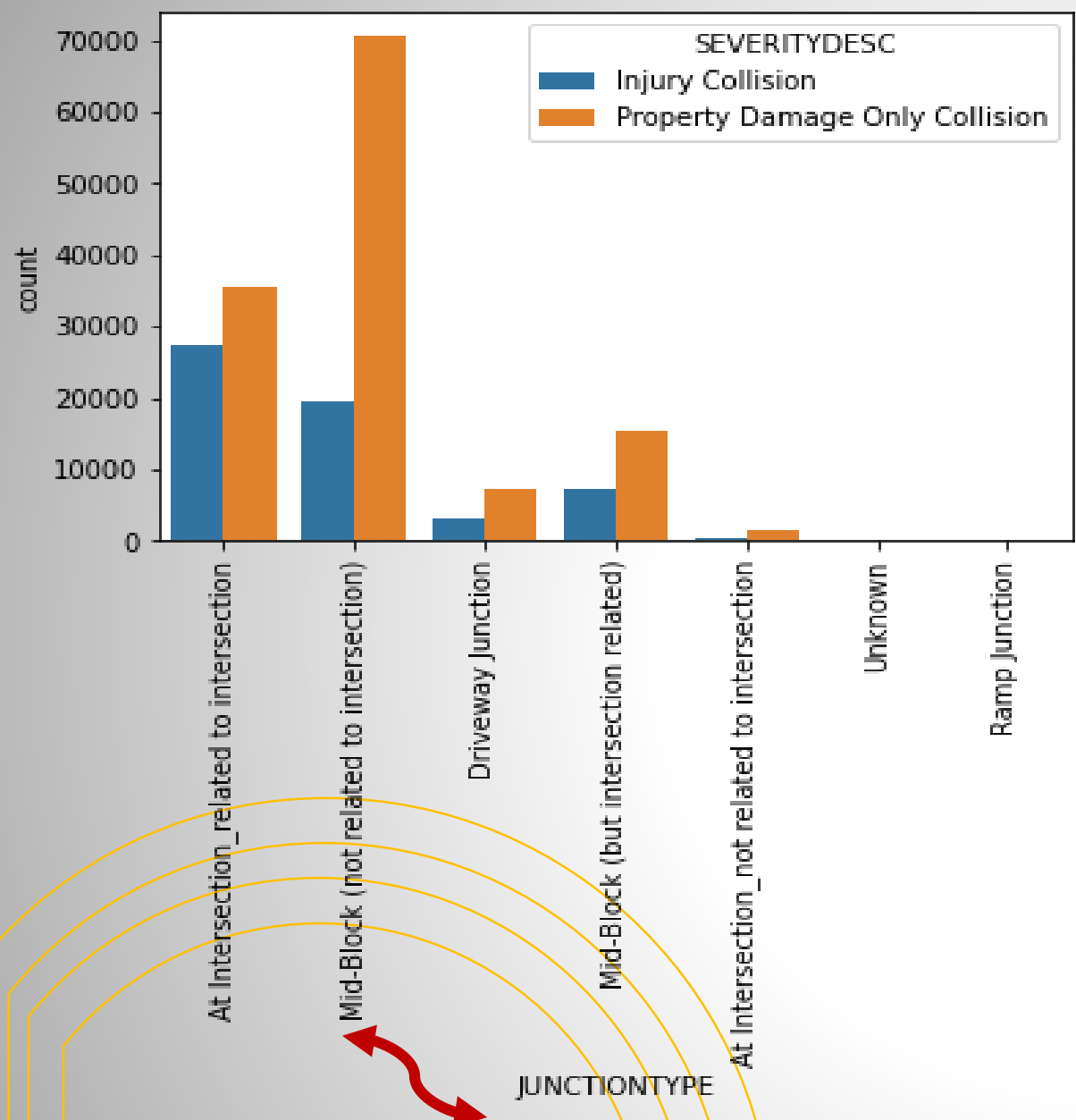


Most of the accidents that occurred involved  
“Property Damage Only Collision”

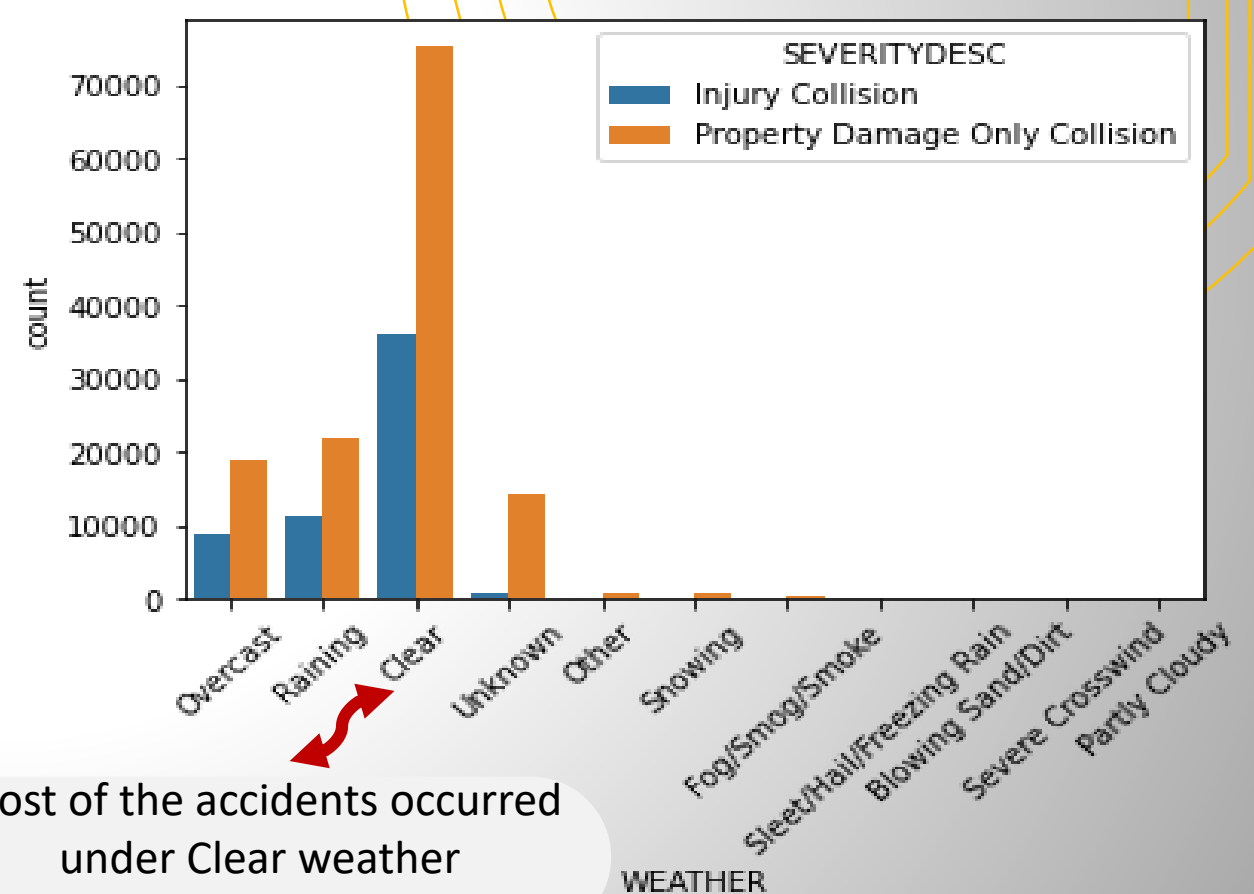


Most of the accidents happened under the Dry road conditions

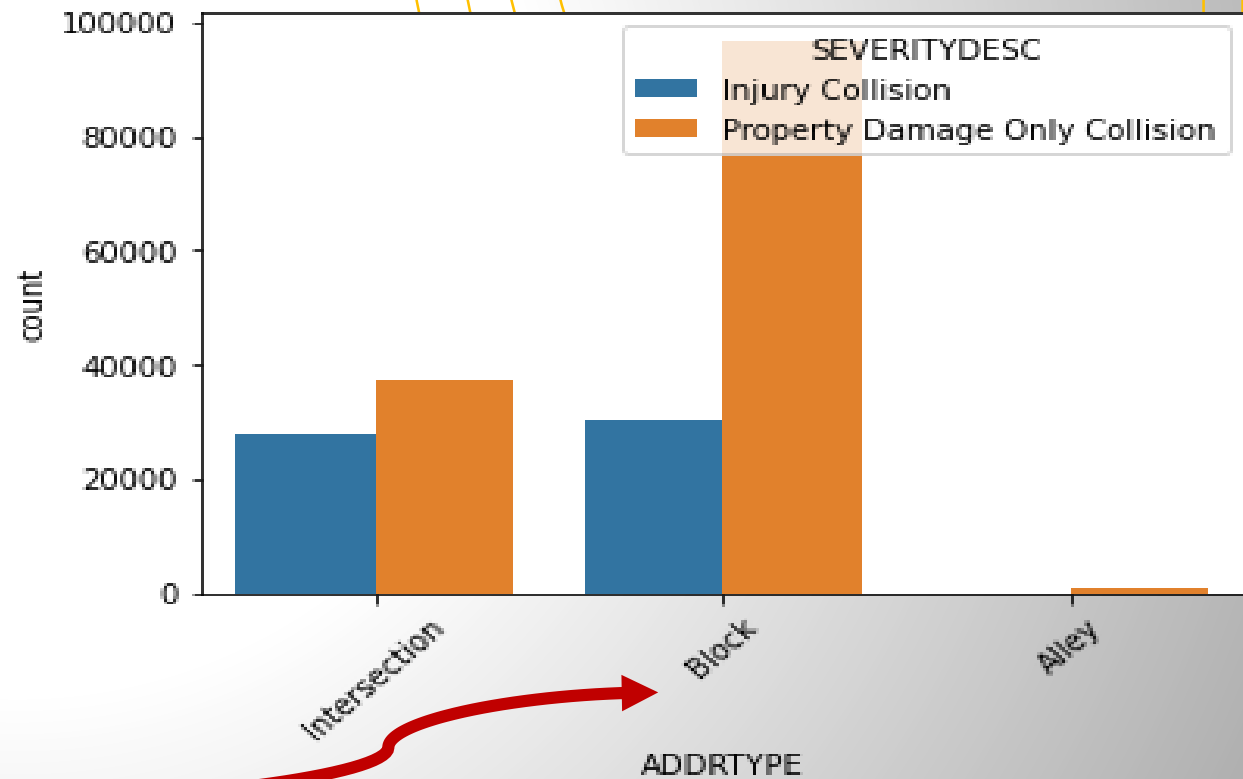
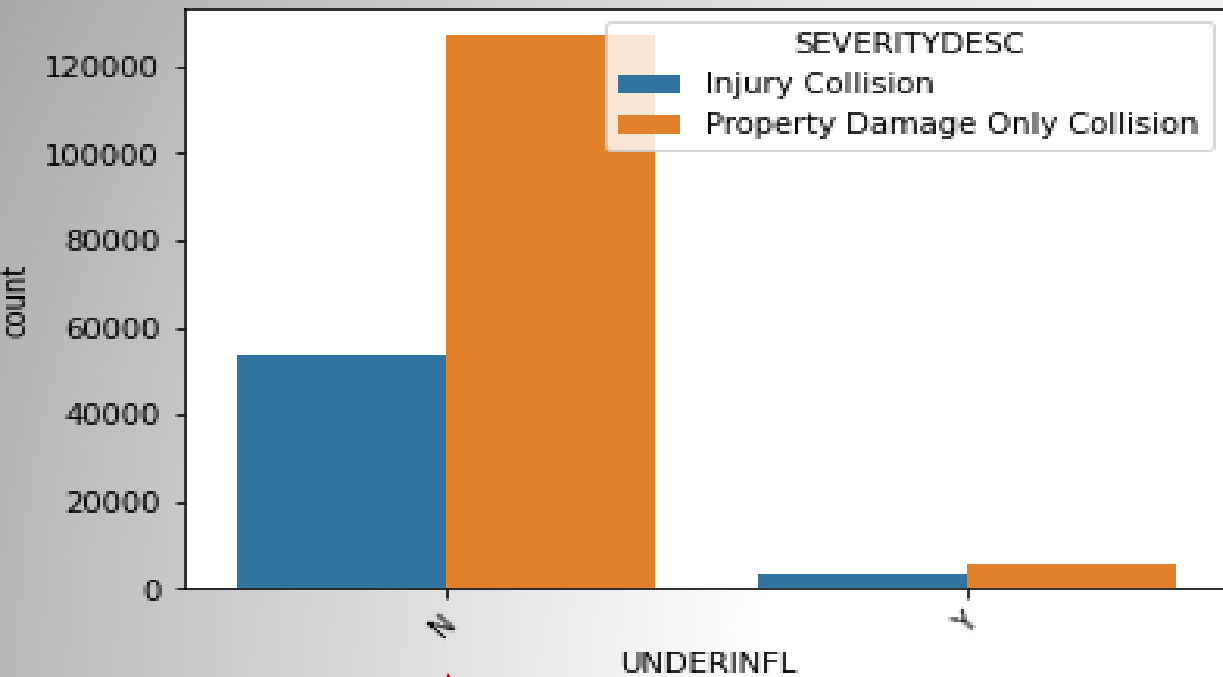




The major junction type involved was "Mid-Block"



Most of the accidents occurred under Clear weather conditions.



Most of the accidents occurred without the “Under influence of Alcohol/Drugs” factor

Property damage only collision has the major address type as the “Block”





### 3. Data Preparation & Modeling

*\*After cleaning the data, features selected for Machine Learning application are mentioned in the following slide*



# Selected Features for applying Machine Learning Algorithms

# "SEVERITYCODE"

## The Target Class(Variable)



- 
- | Feature ID | Feature Name  |
|------------|---------------|
| 1          | longitude     |
| 2          | latitude      |
| 3          | PERSONCOUNT   |
| 4          | PEDCOUNT      |
| 5          | WEATHER       |
| 6          | COLLISIONTYPE |
| 7          | LIGHTCOND     |
| 8          | ROADCOND      |
| 9          | ADDRTYPE      |
| 10         | SDOT_COLDESC  |
| 11         | HITPARKEDCAR  |
| 12         | VEHCOUNT      |
| 13         | PEDCYLCOUNT   |
| 14         | Hour          |

# Machine Learning Algorithms Selection

**1**

**Logistic Regression**

**2**

**K-NN Neighbors**

**3**

**Decision Tree Algorithm**

**4**

**Random Forest Algorithm**

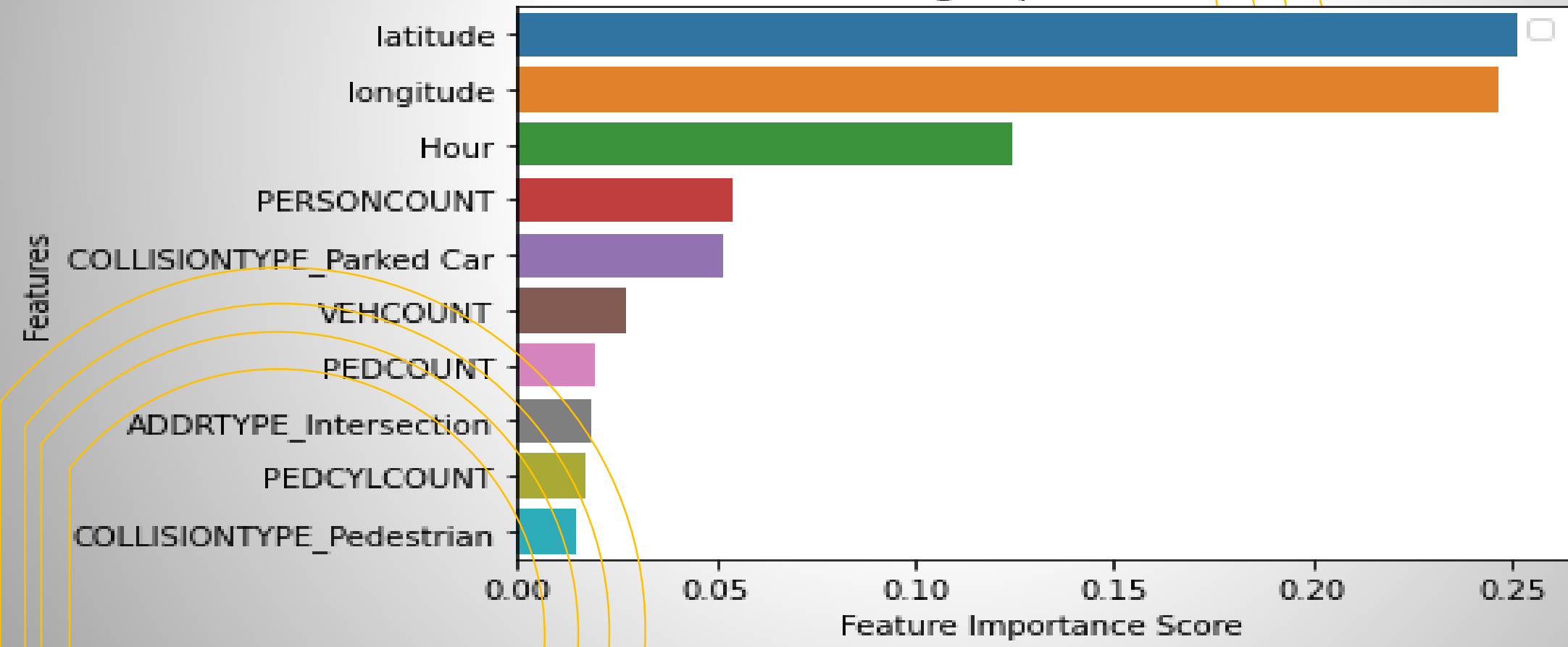


## 4. Evaluation

# FEATURES IMPORTANCE SCORE

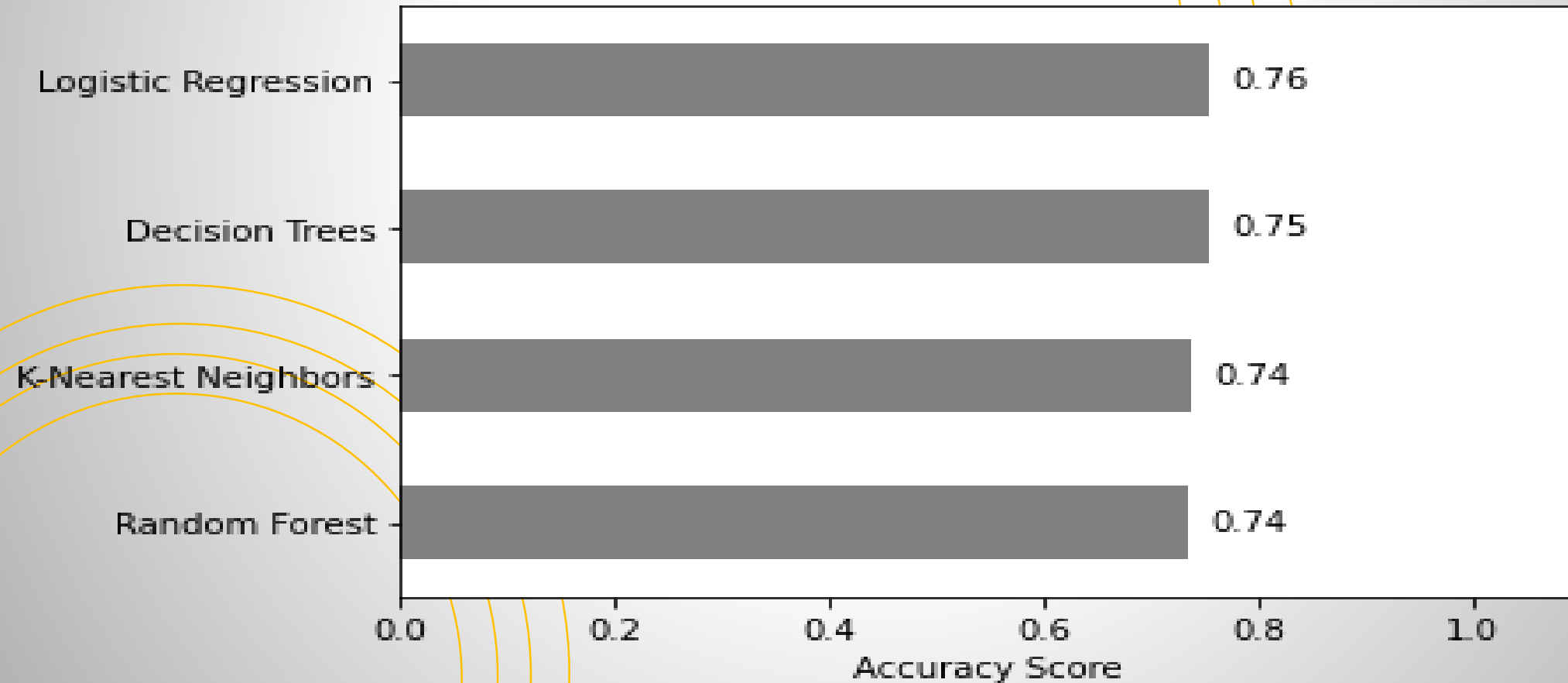


Visualizing Important Features





### Accuracy Score of each Algorithm



## DEPLOYMENT

For the deployment phase as it can vary from project to project a simple pdf report has been generated.

## SUMMARY

- Seattle road accidents data has been analysed in order to get useful insights.
- The data contains multiple attributes e.g. accident severity, collision type, coordinates of the incident, date and time of the incident, weather and road conditions, address types, no of persons injured and property damage and many other attributes.
- There are two accident severity types mentioned in the dataset i.e.
  - Property damage only collision(1)
  - Injury collision(2)
- All the mandatory Cross-industry standard process for data mining CRISP-DM phases are covered in this report which contains the following:
  - Business Understanding
  - Data Understanding
  - Data Preparation
  - Modelling
  - Evaluation
  - Deployment
- In the Modelling phase, four algorithms were selected where the target class was “accident severity”.
- Based on the predictions, “Logistic Regression” relatively performed better among the others having the accuracy percentage of approx.76%.

## CONCLUSION

Based on the selected dataset(features) for this capstone project which includes mainly, coordinates, hour, person count and the collision type, it can be concluded that these particular classes have a somewhat impact on whether or not travelling along the Seattle roads could result in property damage (class 1) or injury (class 2). In this study, the technique of association rules with a large set of accident data to identify the reasons of road accidents were used. The results show that this model could provide good predictions against traffic accident with approx. 76% correct rate. It should be noted that due to the constraints of data and research condition, there are still some factors, such as engine capacity, traffic flows, gender, age of the driver, attaining the missing data etc. that are not considered in this model and can be taken into account for future study. The results of this study can be used in vehicle safety assistance driving and provide early warnings and proposals for safe driving, hence help in reducing the number of accidents.



# Thank You