Predicting Severity Codes of Car Collision Accidents

How we can predict the severity of car collision accidents Using machine learning algorithms

Background / Business Understanding

- Car collisions have been huge problems to our society thus it is important to analyze the historical data and predict before they happen
- Our primary goal is to build an appropriate machine learning model and predict the severity code, which is categorized as follows:
 - 0: Unknown
 - 1 : Property Damage
 - 2 : Injury
 - 2b : Serious Injury
 - 3 : Fatality

Data Acquisition

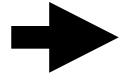
- The data of all collisions in Seattle from 2004 to present has been used
- The raw data have 35 attributes in total including SEVERITYCODE and I chose to have only 15 attributes for further data preparation
- Following is the first five rows of the data:

| | SEVERITYCODE | ADDRTYPE | COLLISIONTYPE | PERSONCOUNT | PEDCOUNT | PEDCYLCOUNT | VEHCOUNT | INCDATE | INATTENTIONIND | UNDERINFL | WEATHER |
|---|--------------|--------------|---------------|-------------|----------|-------------|----------|---------------------------|----------------|-----------|----------|
| 0 | 2 | Intersection | Angles | 2 | 0 | 0 | 2 | 2013/03/27 00:00:00+00 | NaN | N | Overcast |
| 1 | 1 | Block | Sideswipe | 2 | 0 | 0 | 2 | 2006/12/20 00:00:00+00 | NaN | 0 | Raining |
| 2 | 1 | Block | Parked Car | 4 | 0 | 0 | 3 | 2004/11/18 00:00:00+00 | NaN | 0 | Overcast |
| 3 | 1 | Block | Other | 3 | 0 | 0 | 3 | 2013/03/29 00:00:00+00 | NaN | N | Clear |
| 4 | 2 | Intersection | Angles | 2 | 0 | 0 | 2 | 2004/01/28 00:00:00+00 | NaN | 0 | Raining |

Data Preparation

- Balance the data by downsampling
 - As the classes of SEVERITYCODE has different size we need resampling:

```
1 136485
2 58188
Name: SEVERITYCODE, dtype: int64
```



2 58188 1 58188 Name: SEVERITYCODE, dtype: int64

- Fill up the missing values and convert data types
- Create the feature dataset X and target y
 - X: ADDRTYPE, WEATHER, ROADCOND, LIGHTCOND, PERSONCOUNT, PEDCOUNT
 - y: SEVERITYCODE
- Normalize the feature dataset and split the data into train/test set

Methodology / Evaluation

- Three classification algorithms K-Nearest Neighbors, Decision Tree, and Logistic Regression - has been used
- To compare the accuracy of each model, I have calculated f1 score and jaccard score (and log loss only for logistic regression regression)
- All three models give the accuracy between 0.63 and 0.66 and among those models the decision tree algorithm reached the best accuracy

| | KNN | Decision Tree | Logistic Regression |
|---------------|----------|----------------------|----------------------------|
| F1 score | 0.643094 | 0.651396 | 0.633458 |
| Jaccard score | 0.643094 | 0.651396 | 0.633415 |
| Log Loss | NA | NA | 0.635921 |

Conclusion

- All three models yielded the accuracy over 0.6 and among them the decision tree model performed best with the accuracy of 0.652413
- There are several things to consider to improve this modeling:
 - How to choose appropriate attributes and deal with the missing values
 - The characteristics and usages of algorithms