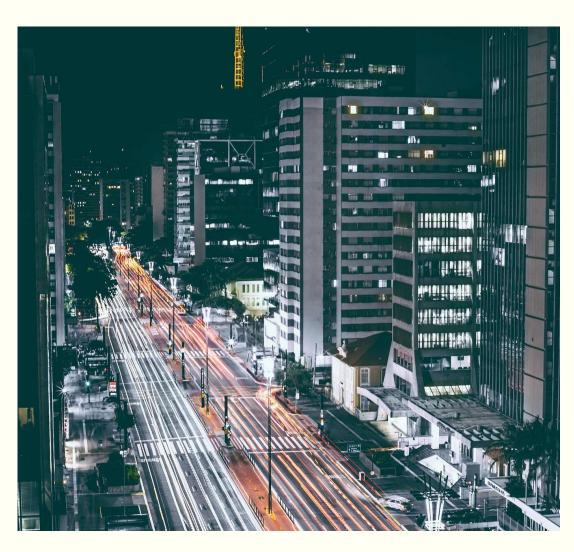
COURSERA CAPSTONE PROJECT

Car Accident Severity

By Michael Fretwell 1 September 2020



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Problem and Background

- According to the Association for Safe International Road Travel (ASIRD)
 approximately 38,000 people die in road crashes in the United States each year.
- The fatality rate is 12.4 deaths per 100,000 inhabitants.
- An additional 4.4 million people are injured seriously enough to require medical attention.
- Road crashes are the leading cause of death in the U.S. for people aged 1-54.
- The ability to leverage available data to make predictions related to automobile accidents would be valuable to a wide range of industry groups.

Scope of Project

- Identification of factors (such as speed, time of day, and weather conditions) that contribute to accidents.
- Build regression classification models to predict severity of accidents with out of sample data.
- Use of regression classification modeling provides for measures of probability, given various conditions, that can be considered as risk of accidents.
- This information will be useful to those in the traffic safety industry for the purpose of implementing actions to reduce risks of accidents.
- Information will also be of use to the insurance industry.

Dataset Information

- Source data is the Seattle "Collisions All Years" dataset produced by the Seattle SDOT Traffic Management Division, Traffic Records Group.
- Point of contact for the data is the SDOT GIS Analyst at <u>DOT_IT_GIS@seattle.gov</u>.
- Period covered by the data is January 2004 to 19 May 2020.
- There are 194,673 observations and 37 attributes.

Dataset Preprocessing

Duplicate or unnecessary features were removed.

Χ	Υ	INCKEY
COLDETKEY	REPORTNO	STATUS
SDOTCOLNUM	LOCATION	SEVERITYCODE.1
SEVERITYDESC	ST COLDESC	EXCEPTRSNCODE
EXCEPTRSNDESC	ST COLCODE	SDOT COLCODE
SDOT COLDESC	INTKEY	INCDATE
OBJECTID	SEGLANEKEY	CROSSWALKKEY

- The label, or target, for models is "SEVERITYCODE" which indicates the severity of the collision ("1" for property damage or "2" for injury).
- Label contains unbalanced data which required balancing to preclude bias in the model.

- Attributes with missing values were identified.
- Analysis of each attribute was done to determine appropriate steps to resolve missing information.
- Replaced missing information with the most frequently occurring value.
- Replaced null values with "0".

```
In [185]: #Look for Missing Values.
          df.isnull().sum()
Out[185]: SEVERITYCODE
                                   0
           ADDRTYPE
                                1926
           COLLISIONTYPE
                                4904
           PERSONCOUNT
           PEDCOUNT
           PEDCYLCOUNT
          VEHCOUNT
           INCDTTM
           JUNCTIONTYPE
                                6329
           INATTENTIONIND
                              164868
           UNDERINFL
                                4884
           WEATHER
                                5081
                                5012
           ROADCOND
           LIGHTCOND
                                5170
           PEDROWNOTGRNT
                              190006
           SPEEDING
                              185340
           HITPARKEDCAR
           dtype: int64
```

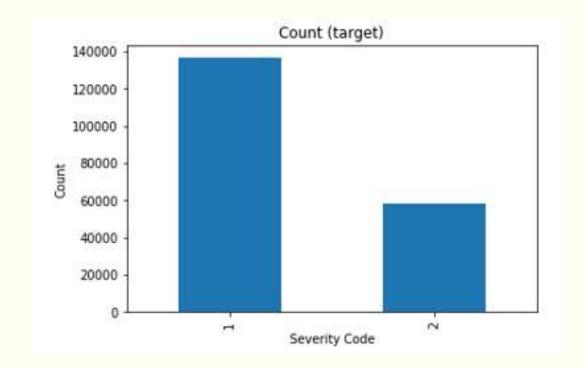
 Categorical features were converted to numbers.

```
OCACHE.
In [218]: #COLLISIONTYPE
          df['COLLISIONTYPE'].value_counts()
Out[210]: Parked Car
                        52891
          Angles
                        34674
          Rear Ended
                       34898
          Other
                        23703
          Sideswipe
                       18609
          Left Turn
                       13703
          Pedestrian
                        6608
          Cycles
                         5415
          Right Turn
                        2956
                         2024
          Head On
          Name: COLLISIONTYPE, dtype: int64
In [211]: df['COLLISIONTYPE'].replace(to_replace=['Parked Car', 'Angles', 'Rear Ended', 'Other', 'Sideswipe',\
                                                 'Left Turn', 'Pedestrian', 'Cycles', 'Right Turn', 'Head On',\
                                                ], value=[0,1,2,3,4,5,6,7,8,9], inplace=True)
```

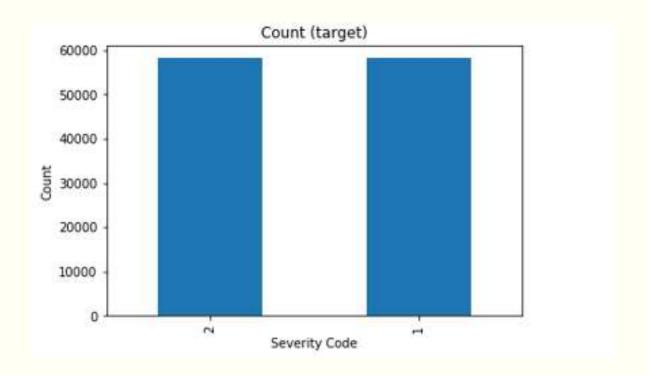
 Object data types were converted to integers.

```
In [223]: #Check Data Types.
          df.dtypes
Out[223]: SEVERITYCODE
                              int64
          ADDRTYPE
                              int64
          COLLISIONTYPE
                              int64
                              int64
          PERSONCOUNT
          PEDCOUNT
                              int64
          PEDCYLCOUNT
                              int64
          VEHCOUNT
                              int64
                             object
          INCDTTM
          JUNCTIONTYPE
                              int64
          INATTENTIONIND
                              int64
                             object
          UNDERINFL
          WEATHER
                              int64
          ROADCOND
                              int64
          LIGHTCOND
                              int64
          PEDROWNOTGRNT
                              int64
          SPEEDING
                              int64
                             object
          HITPARKEDCAR
          dtype: object
In [224]: #Convert Objects to Integers.
          df['UNDERINFL'] = pd.to_numeric(df['UNDERINFL'])
          df['WEATHER'] = pd.to_numeric(df['WEATHER'])
          df['HITPARKEDCAR'] = pd.to_numeric(df['HITPARKEDCAR'])
```

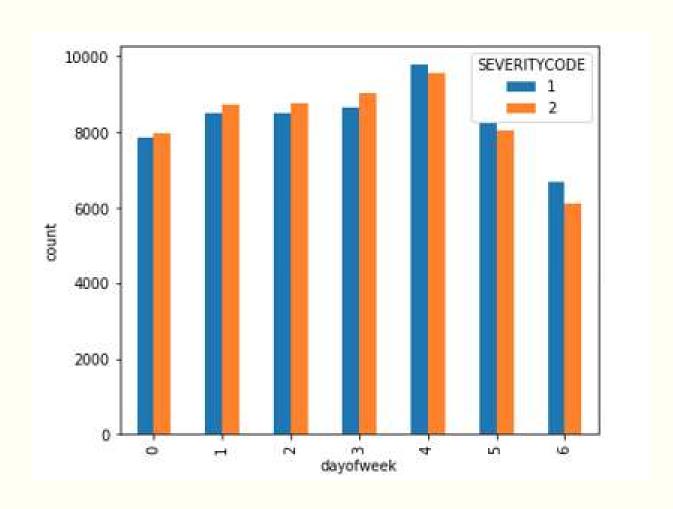
- Target for models was Severity Code.
 - 1 Property Damage
 - 2 Injury
- Target data was unbalanced.
- Would create bias in models if not adjusted.



- Steps were taken to balance the dataset.
- Balanced dataset consists of 116,376 records.

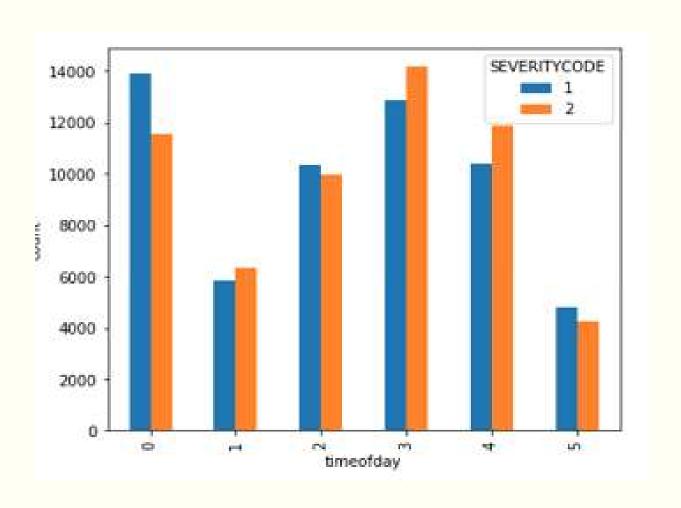


- Incident date time feature (INCDTTM) was an object.
- Conversion to date time format allowed for calculations.
- Day of week analysis shows mostly minor variations in accidents by day.
- Noticeable drop in number of accidents on Saturdays.



- Time of day grouped into 6 bins for analysis.
- Fewer number of accidents during the periods 0400-0800 and 2000-0000.

Time of Day	Bin	
0000 to 0400	0	
0400 to 0800	1	
0800 to 1200	2	
1200 to 1600	3	
1600 to 2000	4	
2000 to 0000	5	



- 15 features were selected for use in the models from the initial attributes.
- 2 additional features were derived from the INCDTTM field:
 - dayofweek
 - timeofday

ADDRTYPE	COLLISIONTYPE	PERSONCOUNT
PEDCOUNT	PEDCYLCOUNT	VEHCOUNT
JUNCTIONTYPE	INATTENTIONIND	UNDERINFL
WEATHER	ROADCOND	LIGHTCOND
PEDROWNOTGRNT	SPEEDING	HITPARKEDCAR
dayofweek	timeofday	

Modeling

- Objective of models is to predict severity of accidents; thus Classification models were used.
- K Nearest Neighbor
- Decision Tree
- Support Vector Machine
- Logistic Regression

9/1/2020 15

Modeling (Cont.)

- Since all data was manually converted to numerical, uncertain if sklearn preprocessing to further standardize data would have an impact.
- Models were run with and without preprocessing variances were minor.
- Data was split into Train and Test groups
 - Train set: (93,100, 17) (93,100,)
 - Test set: (23,273, 17) (23,276,)

Model Evaluation

- Accuracy of models was evaluated using sklearn metrics library:
 - Jaccard
 - F-1
 - LogLoss
- Models were run with and without sklearn preprocessing resulting in minor differences.
- In both cases, decision tree and logistic regression models produced slightly better scores.

	Jaccard	F1-Score	LogLoss
Algorithm			
KNN	0.66	0.68	17.31
Decision Tree	0.71	0.68	17.31
SVM	0.66	0.68	17.31
Logistic Regression	0.71	0.68	17.31

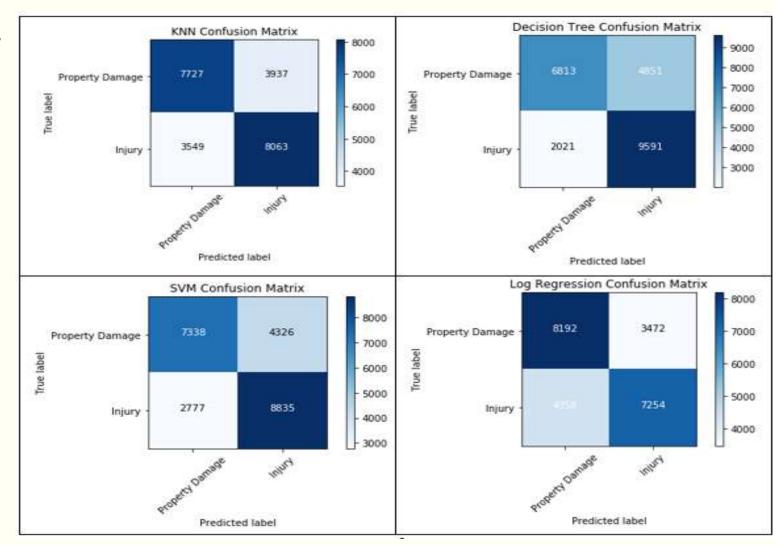
Scores Without Preprocessing

	Jaccard	F1-Score	LogLoss
Algorithm			
KNN	0.68	0.67	17.31
Decision Tree	0.71	0.67	17.31
SVM	0.69	0.68	17.31
Logistic Regression	0.71	0.67	17.31

Scores With Preprocessing

Model Evaluation (Cont.)

- Confusion matrices for each model.
- Confirm decision tree and logistic regression best results.
- Decision tree better predicting accidents with injuries.
- Logistics Regression better predicting accidents with property damage.



Conclusion

- Using Seattle accident information dataset we processed the data and built classification models that predicting accident severity with reasonable accuracy.
- We noted a noticeable drop in number of accidents occurring on Saturdays.
- We also noted significantly fewer accidents occurred during the time periods 0400-0800 and 2000-0000.
- There is room for improvement in accuracy of models.
- Future efforts could perform more feature evaluation to determine best features for use in models.
- Additional datasets could be added to include information such as nationwide data and population information.
- Categorical target data could be expanded to additional classifications such as fatality information.