Accident Severity Prediction

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Introduction

- In an effort to reduce the frequency of car collisions in a community, an algorithm must be developed to predict the severity of an accident given the current weather, road conditions , visibility conditions and also based on the day of week(weekday or weekend).
- When conditions are bad, this model will alert drivers to remind them to be more careful.

Data

- The data is the accident data for Great Britain , dated from 2005 to 2010 . This dataset was available on kaggle :- https://www.kaggle.com/pachriisk/great-britain-road-accidents?select=uk-traffic-accidents-columns.csv
- The different columns of the dataset like light conditions , road_surface_conditions , weather_conditions , day_of_week will help us to predict the feature - accident severity .

Contents of Dataset

1. Accident_Index
2. Accident_Severity
3. Date
4. Day_of_Week
5. Light_Conditions
6. Number_of_Casualties
7. Number_of_Vehicles
8. Pedestrian_Crossing-Human_Control
9. Pedestrian_Crossing-Physical_Facilities
10. Road_Surface_Conditions
11. Speed_limit
12. Time
13. Weather_Conditions
14. Year

Methodology

- I have used the KNN classification for predicting the severity of the accident using the features like road conditions , weather conditions , light conditions and day of week.

- Firstly we preprocess the data by dropping the irrelevant columns , dropping the null values and encoding the data.

- Then we use the train_test_split for splitting the data for training and testing.

Methodology

- Creating the model:
  Firstly , we try to create the knn model using the k=6 , we train the model using fit_transform and predict some values from it.

- Finding the accuracy:
  Now , we try to find the accuracy of the model by running it on the test data , and then finding the optimal K value for our classifier.

In [5]:
```
# Dropping the unwanted columns
df=df.drop(['Pedestrian_Crossing-Human_Control','Pedestrian_Crossing-Physical_Facilities','Accident_Index'],axis=1)
```

In [6]:
```
df.columns
```

Out[6]:
```
Index(['Accident_Severity', 'Date', 'Day_of_Week', 'Light_Conditions',
       'Number_of_Casualties', 'Number_of_Vehicles', 'Road_Surface_Conditions',
       'Speed_limit', 'Time', 'Weather_Conditions', 'Year'],
      dtype='object')
```

In [7]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1048575 entries, 0 to 1048574
Data columns (total 11 columns):
 #   Column                   Non-Null Count    Dtype
---  ------                   --------------    -----
 0   Accident_Severity        1048575 non-null  object
 1   Date                     1048575 non-null  object
 2   Day_of_Week              1048575 non-null  object
 3   Light_Conditions         1048575 non-null  object
 4   Number_of_Casualties     1048575 non-null  int64
 5   Number_of_Vehicles       1048575 non-null  int64
 6   Road_Surface_Conditions  1048575 non-null  object
 7   Speed_limit              1048575 non-null  int64
 8   Time                     1048475 non-null  object
 9   Weather_Conditions       1048575 non-null  object
 10  Year                     1048575 non-null  int64
dtypes: int64(4), object(7)
memory usage: 60.0+ MB
```

In [8]:
```
# Dropping the null values
df=df.dropna()
```

Methodology

In [16]:
```
# We need to encode the data for applying KNN classifier on it
# Using label encoder for it.
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
```

In [17]:
```
# Encoding the day
day=le.fit_transform(df['Day_of_Week'])
```

In [18]:
```
# Encoding the light conditions
light=le.fit_transform(df['Light_Conditions'])
```

In [19]:
```
# Encoding the road surface conditions
road=le.fit_transform(df['Road_Surface_Conditions'])
```

In [20]:
```
# Encoding the weather conditions
weather=le.fit_transform(df['Weather_Conditions'])
```

In [21]:
```
# Encoding the accident severity
severity=le.fit_transform(df['Accident_Severity'])
```

In [22]:
```
# Zipping all the features needed for classification
features=list(zip(day,light,road,weather))
```

In [23]:
```
from sklearn.model_selection import train_test_split
```

In [24]:
```
# splitting the dataset for training and testing.
xtrain, xtest, ytrain, ytest = train_test_split(features,severity, test_size=0.6)
```

Methodology

In [37]:
```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score
import matplotlib.pyplot as plt
```

In [26]:
```
# Using the classifier for k=6 neighbors
model = KNeighborsClassifier(n_neighbors=6)
model.fit(xtrain,ytrain)
```

Out[26]:
```
KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
                     metric_params=None, n_jobs=None, n_neighbors=6, p=2,
                     weights='uniform')
```

In [27]:
```
# Using the encoded fields for prediction
predicted= model.predict([[0,0,0,0]])
print(predicted)
```

```
[2]
```

Methodology

In [30]:
```
# As the data is big , using only half the data for finding accuracy and the optimal K value.
tt = len(xtrain)
tv = len(xtest)
len(xtrain[int(tt*0.5):]), len(xtest[int(tv*0.5):])
```

Out[30]:
```
(209695, 314543)
```

In [34]:
```
# Finding the best value of K
ks = 10
mean_acc = np.zeros(ks-1)
std_acc = np.zeros(ks-1)

for n in range(4,ks,2):
    neigh = KNeighborsClassifier(n_neighbors = n).fit(xtrain[int(tt*0.5):],ytrain[int(tt*0.5):])
    yhat = neigh.predict(xtest[int(tv*0.5):])
    mean_acc[n-1] = accuracy_score(ytest[int(tv*0.5):],yhat)
    std_acc[n-1] = np.std(yhat==ytest[int(tv*0.5):])/np.sqrt(len(yhat))
print('Best performing K is '+ str(mean_acc.argmax()+1) + ' with an accuracy of ' +str(mean_acc.max()))
```

```
Best performing K is 8 with an accuracy of 0.8491462216612673
```

The best value of K is 8 and it gives an accuracy of 84.9%

Methodology

Results

- We have used KNN classifiers to predict the severity of an accident on the basis of various features . We have found the accuracy of the model which is around 85% and this accuracy is got when k=8.

- Secondly, we have found that the KNN classifier is best for this problem statement because:-
  Very simple implementation. Robust with regard to the search space; for instance, classes don't have to be linearly separable.

Conclusion

Hence , we can predict the severity of any accident on the basis of the various features like light conditions , weather conditions ,etc and hence we can predict how much time the traffic would be stuck on the basis of severity of the accident.

The End