1. 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.

2. Data understanding:

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 .
Accident severity are:-
Slight
Serious
Fatal

All the columns like weather conditions , light conditions , road conditions , severity can be encoded for using a classifier on it.

3. 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.

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()
```

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)
```

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.

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]
```

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 [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%

4.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.

5.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.