

```
import pandas as pd
import matplotlib.pyplot as plt
```

```
import numpy as np
```

```
accidents=pd.read_csv("/content/accidents.csv")
vehicles=pd.read_csv("/content/vehicles.csv")
```

```
accidents.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 88624 entries, 0 to 88623
Data columns (total 32 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   Accident_Index                           88624 non-null  object
1   Location_Easting_OSGR                    88592 non-null  float64
2   Location_Northing_OSGR                   88592 non-null  float64
3   Longitude                                88592 non-null  float64
4   Latitude                                  88592 non-null  float64
5   Police_Force                             88623 non-null  float64
6   Accident_Severity                        88623 non-null  float64
7   Number_of_Vehicles                       88623 non-null  float64
8   Number_of_Casualties                     88623 non-null  float64
9   Date                                     88623 non-null  object
10  Day_of_Week                             88623 non-null  float64
11  Time                                    88622 non-null  object
12  Local_Authority_(District)               88623 non-null  float64
13  Local_Authority_(Highway)                88623 non-null  object
14  1st_Road_Class                           88623 non-null  float64
15  1st_Road_Number                          88623 non-null  float64
16  Road_Type                                88623 non-null  float64
17  Speed_limit                              88623 non-null  float64
18  Junction_Detail                          88623 non-null  float64
19  Junction_Control                         88623 non-null  float64
20  2nd_Road_Class                           88623 non-null  float64
21  2nd_Road_Number                          88623 non-null  float64
22  Pedestrian_Crossing-Human_Control         88623 non-null  float64
23  Pedestrian_Crossing-Physical_Facilities   88623 non-null  float64
24  Light_Conditions                         88623 non-null  float64
25  Weather_Conditions                       88623 non-null  float64
26  Road_Surface_Conditions                   88623 non-null  float64
27  Special_Conditions_at_Site                88623 non-null  float64
28  Carriageway_Hazards                      88623 non-null  float64
29  Urban_or_Rural_Area                      88623 non-null  float64
30  Did_Police_Officer_Attend_Scene_of_Accident 88623 non-null  float64
31  LSOA_of_Accident_Location                88512 non-null  object
dtypes: float64(27), object(5)
memory usage: 21.6+ MB
```

```
vehicles.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 46070 entries, 0 to 46069
Data columns (total 22 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   Accident_Index                           46070 non-null  object
1   Vehicle_Reference                        46069 non-null  float64
2   Vehicle_Type                            46069 non-null  float64
3   Towing_and_Articulation                 46069 non-null  float64
4   Vehicle_Manoeuvre                       46069 non-null  float64
5   Vehicle_Location-Restricted_Lane        46069 non-null  float64
6   Junction_Location                      46069 non-null  float64
7   Skidding_and_Overturning                46069 non-null  float64
8   Hit_Object_in_Carriageway               46069 non-null  float64
9   Vehicle_Leaving_Carriageway             46069 non-null  float64
10  Hit_Object_off_Carriageway              46069 non-null  float64
11  1st_Point_of_Impact                     46069 non-null  float64
12  Was_Vehicle_Left_Hand_Drive?            46069 non-null  float64
13  Journey_Purpose_of_Driver                 46069 non-null  float64
14  Sex_of_Driver                           46069 non-null  float64
15  Age_of_Driver                           46069 non-null  float64
16  Age_Band_of_Driver                      46069 non-null  float64
17  Engine_Capacity_(CC)                    46069 non-null  float64
18  Propulsion_Code                         46069 non-null  float64
19  Age_of_Vehicle                          46069 non-null  float64
20  Driver_IMD_Decile                       46069 non-null  float64
21  Driver_Home_Area_Type                    46069 non-null  float64
dtypes: float64(21), object(1)
memory usage: 7.7+ MB
```

```
accidents["Accident_Severity"].unique()
#gives unique values in the column
```

```
array([ 2.,  3.,  1., nan])
```

```
#Combining two datasets
import pandas as pd
data=pd.merge(accidents,vehicles,on="Accident_Index",how='outer')
```

```
data.head()
```

data.info()

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 109135 entries, 0 to 109134
Data columns (total 53 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   Accident_Index                           109135 non-null  object
1   Location_Easting_OSGR                   109102 non-null  float64
2   Location_Northing_OSGR                  109102 non-null  float64
3   Longitude                               109102 non-null  float64
4   Latitude                                109102 non-null  float64
5   Police_Force                            109133 non-null  float64
6   Accident_Severity                       109133 non-null  float64
7   Number_of_Vehicles                      109133 non-null  float64
8   Number_of_Casualties                    109133 non-null  float64
9   Date                                    109133 non-null  object
10  Day_of_Week                             109133 non-null  float64
11  Time                                    109132 non-null  object
12  Local_Authority_(District)              109133 non-null  float64
13  Local_Authority_(Highway)               109133 non-null  object
14  1st_Road_Class                          109133 non-null  float64
15  1st_Road_Number                         109133 non-null  float64
16  Road_Type                              109133 non-null  float64
17  Speed_limit                             109133 non-null  float64
18  Junction_Detail                         109133 non-null  float64
19  Junction_Control                        109133 non-null  float64
20  2nd_Road_Class                          109133 non-null  float64
21  2nd_Road_Number                         109133 non-null  float64
22  Pedestrian_Crossing-Human_Control        109133 non-null  float64
23  Pedestrian_Crossing-Physical_Facilities  109133 non-null  float64
24  Light_Conditions                        109133 non-null  float64
25  Weather_Conditions                      109133 non-null  float64
26  Road_Surface_Conditions                  109133 non-null  float64
27  Special_Conditions_at_Site               109133 non-null  float64
28  Carriageway_Hazards                     109133 non-null  float64
29  Urban_or_Rural_Area                     109133 non-null  float64
30  Did_Police_Officer_Attend_Scene_of_Accident  109133 non-null  float64
31  LSOA_of_Accident_Location               108975 non-null  object
32  Vehicle_Reference                       46069 non-null  float64
33  Vehicle_Type                            46069 non-null  float64
34  Towing_and_Articulation                 46069 non-null  float64
35  Vehicle_Manoeuvre                       46069 non-null  float64
36  Vehicle_Location-Restricted_Lane         46069 non-null  float64
37  Junction_Location                       46069 non-null  float64
38  Skidding_and_Overturning                46069 non-null  float64
39  Hit_Object_in_Carriageway               46069 non-null  float64
40  Vehicle_Leaving_Carriageway             46069 non-null  float64
41  Hit_Object_off_Carriageway              46069 non-null  float64
42  1st_Point_of_Impact                     46069 non-null  float64
43  Was_Vehicle_Left_Hand_Drive?            46069 non-null  float64
44  Journey_Purpose_of_Driver                 46069 non-null  float64
45  Sex_of_Driver                           46069 non-null  float64
46  Age_of_Driver                           46069 non-null  float64
47  Age_Band_of_Driver                      46069 non-null  float64

```

48	Engine_Capacity_(CC)	46069	non-null	float64
49	Propulsion_Code	46069	non-null	float64
50	Age_of_Vehicle	46069	non-null	float64
51	Driver_IMD_Decile	46069	non-null	float64
52	Driver_Home_Area_Type	46069	non-null	float64

Description of variables:

Accident\_Severity:1-Fatal,2-Serious,3-Slight  
 Road\_Type:1-Roundabout,2-One way street,3-Dual carriageway,  
 6-Single carriageway,7-Slip road,9-Unknown  
 Road\_Surface\_Conditions:1-Dry,2-Wet or damp,3-Snow,4-Frost or ice,  
 5-Flood over 3cm. deep  
 Skidding\_and\_Overturning:0-None,1-Skidded,2-Skidded and overturned,  
 5-Overturned  
 Weather\_Conditions:1-Fine no high winds,2-Raining no high winds,  
 3-Snowing no high winds,4-Fine + high winds,5-Raining + high winds,  
 6-Snowing + high winds,7-Fog or mist,8-Other,9-Unknown

Description of variables: Accident\_Severity:1-Fatal,2-Serious,3-Slight Road\_Type:1-Roundabout,2-One way street,3-Dual carriageway, 6-Single carriageway,7-Slip road,9-Unknown  
 Road\_Surface\_Conditions:1-Dry,2-Wet or damp,3-Snow,4-Frost or ice, 5-Flood over 3cm. deep  
 Skidding\_and\_Overturning:0-None,1-Skidded,2-Skidded and overturned, 5-Overturned  
 Weather\_Conditions:1-Fine no high winds,2-Raining no high winds, 3-Snowing no high winds,4-Fine + high winds,5-Raining + high winds, 6-Snowing + high winds,7-Fog or mist,8-Other,9-Unknown

+ Code

+ Text

```
#Filling the missing values by using dropna() method.
#Our dataset contains null values in the format -1 and NAN
for column in data.columns:
    data[data[column]==-1]
data.dropna(inplace=True)
```

```
data.isnull().sum()
#We can see that all columns have no null values.
```

Accident_Index	0
Location_Easting_OSGR	0
Location_Northing_OSGR	0
Longitude	0
Latitude	0
Police_Force	0
Accident_Severity	0
Number_of_Vehicles	0
Number_of_Casualties	0
Date	0
Day_of_Week	0
Time	0
Local_Authority_(District)	0
Local_Authority_(Highway)	0
1st_Road_Class	0
1st_Road_Number	0
Road_Type	0
Speed_limit	0

```

Junction_Detail                                0
Junction_Control                              0
2nd_Road_Class                                0
2nd_Road_Number                              0
Pedestrian_Crossing-Human_Control             0
Pedestrian_Crossing-Physical_Facilities       0
Light_Conditions                             0
Weather_Conditions                           0
Road_Surface_Conditions                      0
Special_Conditions_at_Site                   0
Carriageway_Hazards                          0
Urban_or_Rural_Area                          0
Did_Police_Officer_Attend_Scene_of_Accident  0
LSOA_of_Accident_Location                    0
Vehicle_Reference                            0
Vehicle_Type                                  0
Towing_and_Articulation                      0
Vehicle_Manoeuvre                            0
Vehicle_Location-Restricted_Lane             0
Junction_Location                           0
Skidding_and_Overturning                     0
Hit_Object_in_Carriageway                    0
Vehicle_Leaving_Carriageway                  0
Hit_Object_off_Carriageway                   0
1st_Point_of_Impact                          0
Was_Vehicle_Left_Hand_Drive?                 0
Journey_Purpose_of_Driver                      0
Sex_of_Driver                                0
Age_of_Driver                                0
Age_Band_of_Driver                           0
Engine_Capacity_(CC)                         0
Propulsion_Code                              0
Age_of_Vehicle                               0
Driver_IMD_Decile                             0
Driver_Home_Area_Type                         0
dtype: int64

```

```
data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 14027 entries, 9 to 46068
Data columns (total 53 columns):

```

#	Column	Non-Null Count	Dtype
0	Accident_Index	14027 non-null	object
1	Location_Easting_OSGR	14027 non-null	float64
2	Location_Northing_OSGR	14027 non-null	float64
3	Longitude	14027 non-null	float64
4	Latitude	14027 non-null	float64
5	Police_Force	14027 non-null	float64
6	Accident_Severity	14027 non-null	float64
7	Number_of_Vehicles	14027 non-null	float64
8	Number_of_Casualties	14027 non-null	float64
9	Date	14027 non-null	object
10	Day_of_Week	14027 non-null	float64

11	Time	14027	non-null	object
12	Local_Authority_(District)	14027	non-null	float64
13	Local_Authority_(Highway)	14027	non-null	object
14	1st_Road_Class	14027	non-null	float64
15	1st_Road_Number	14027	non-null	float64
16	Road_Type	14027	non-null	float64
17	Speed_limit	14027	non-null	float64
18	Junction_Detail	14027	non-null	float64
19	Junction_Control	14027	non-null	float64
20	2nd_Road_Class	14027	non-null	float64
21	2nd_Road_Number	14027	non-null	float64
22	Pedestrian_Crossing-Human_Control	14027	non-null	float64
23	Pedestrian_Crossing-Physical_Facilities	14027	non-null	float64
24	Light_Conditions	14027	non-null	float64
25	Weather_Conditions	14027	non-null	float64
26	Road_Surface_Conditions	14027	non-null	float64
27	Special_Conditions_at_Site	14027	non-null	float64
28	Carriageway_Hazards	14027	non-null	float64
29	Urban_or_Rural_Area	14027	non-null	float64
30	Did_Police_Officer_Attend_Scene_of_Accident	14027	non-null	float64
31	LSOA_of_Accident_Location	14027	non-null	object
32	Vehicle_Reference	14027	non-null	float64
33	Vehicle_Type	14027	non-null	float64
34	Towing_and_Articulation	14027	non-null	float64
35	Vehicle_Manoeuvre	14027	non-null	float64
36	Vehicle_Location-Restricted_Lane	14027	non-null	float64
37	Junction_Location	14027	non-null	float64
38	Skidding_and_Overturning	14027	non-null	float64
39	Hit_Object_in_Carriageway	14027	non-null	float64
40	Vehicle_Leaving_Carriageway	14027	non-null	float64
41	Hit_Object_off_Carriageway	14027	non-null	float64
42	1st_Point_of_Impact	14027	non-null	float64
43	Was_Vehicle_Left_Hand_Drive?	14027	non-null	float64
44	Journey_Purpose_of_Driver	14027	non-null	float64
45	Sex_of_Driver	14027	non-null	float64
46	Age_of_Driver	14027	non-null	float64
47	Age_Band_of_Driver	14027	non-null	float64
48	Engine_Capacity_(CC)	14027	non-null	float64
49	Propulsion_Code	14027	non-null	float64
50	Age_of_Vehicle	14027	non-null	float64
51	Driver_IMD_Decile	14027	non-null	float64
52	Driver_Home_Area_Type	14027	non-null	float64

```
data['Accident_Severity'].unique()
```

```
#The column 'Accident_Severity' has 3 unique values:1,2,3 indicating Fatal,Severe and Slight Accident
```

```
array([3., 2., 1.])
```

```
data['Road_Type'].unique()
```

```
array([6., 2., 3., 1., 9., 7.])
```

```
data['Road_Surface_Conditions'].unique()
```

```
array([1., 2., 4., 3., 5.])
```

```
data['Skidding_and_Overturning'].unique()
```

```
array([0., 1., 5., 2.])
```

```
import pandas as pd
data.to_csv("Accidents_data.csv")
```

```
data['Accident_Severity'].unique()
```

```
array([3., 2., 1.])
```

```
data['Weather_Conditions'].unique()
```

```
array([1., 2., 3., 8., 4., 5., 9., 7., 6.])
```

```
data[(data['Accident_Severity']!=1)].count()
```

Accident_Index	13957
Location_Easting_OSGR	13957
Location_Northing_OSGR	13957
Longitude	13957
Latitude	13957
Police_Force	13957
Accident_Severity	13957
Number_of_Vehicles	13957
Number_of_Casualties	13957
Date	13957
Day_of_Week	13957
Time	13957
Local_Authority_(District)	13957
Local_Authority_(Highway)	13957
1st_Road_Class	13957
1st_Road_Number	13957
Road_Type	13957
Speed_limit	13957
Junction_Detail	13957
Junction_Control	13957
2nd_Road_Class	13957
2nd_Road_Number	13957
Pedestrian_Crossing-Human_Control	13957
Pedestrian_Crossing-Physical_Facilities	13957
Light_Conditions	13957
Weather_Conditions	13957
Road_Surface_Conditions	13957
Special_Conditions_at_Site	13957
Carriageway_Hazards	13957
Urban_or_Rural_Area	13957
Did_Police_Officer_Attend_Scene_of_Accident	13957
LSOA_of_Accident_Location	13957
Vehicle_Reference	13957
Vehicle_Type	13957
Towing_and_Articulation	13957

```

Vehicle_Manoeuvre 13957
Vehicle_Location-Restricted_Lane 13957
Junction_Location 13957
Skidding_and_Overturning 13957
Hit_Object_in_Carriageway 13957
Vehicle_Leaving_Carriageway 13957
Hit_Object_off_Carriageway 13957
1st_Point_of_Impact 13957
Was_Vehicle_Left_Hand_Drive? 13957
Journey_Purpose_of_Driver 13957
Sex_of_Driver 13957
Age_of_Driver 13957
Age_Band_of_Driver 13957
Engine_Capacity_(CC) 13957
Propulsion_Code 13957
Age_of_Vehicle 13957
Driver_IMD_Decile 13957
Driver_Home_Area_Type 13957
dtype: int64

```

Visualization:

How many accidents are severe to the most?

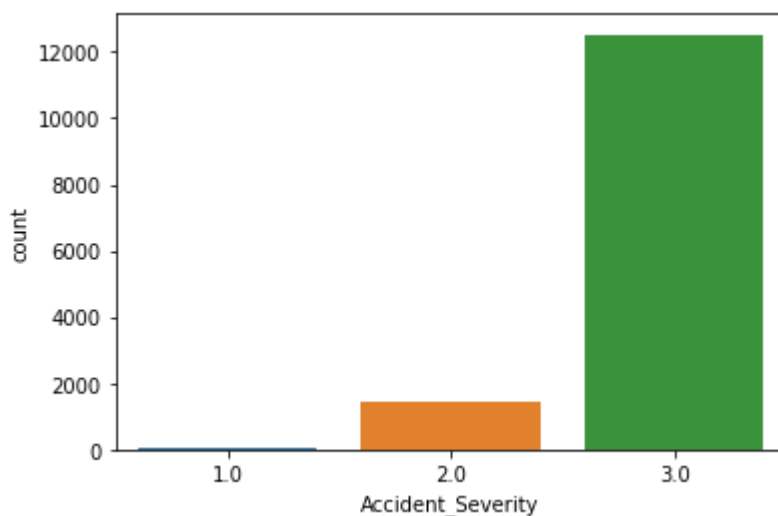
```
import seaborn as sns
```

```
sns.countplot(data['Accident_Severity'])
```

```

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword arguments: {'x': 'Accident_Severity', 'y': 'count'}. This warning will be removed in a future version of Seaborn.
FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7f94081ed110>

```



From the plot, we can see that most of the accidents are slight only.



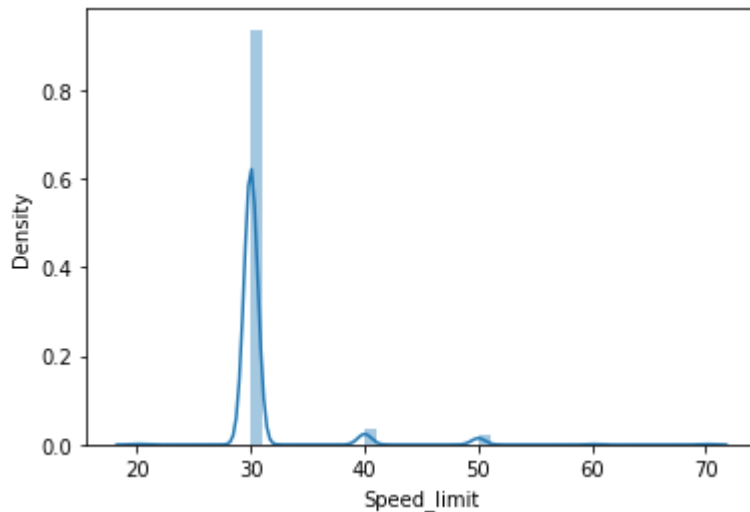
```
data[data['Accident_Severity']==1].count()
```

Accident_Index	70
Location_Easting_OSGR	70
Location_Northing_OSGR	70
Longitude	70
Latitude	70
Police_Force	70
Accident_Severity	70
Number_of_Vehicles	70
Number_of_Casualties	70
Date	70
Day_of_Week	70
Time	70
Local_Authority_(District)	70
Local_Authority_(Highway)	70
1st_Road_Class	70
1st_Road_Number	70
Road_Type	70
Speed_limit	70
Junction_Detail	70
Junction_Control	70
2nd_Road_Class	70
2nd_Road_Number	70
Pedestrian_Crossing-Human_Control	70
Pedestrian_Crossing-Physical_Facilities	70
Light_Conditions	70
Weather_Conditions	70
Road_Surface_Conditions	70
Special_Conditions_at_Site	70
Carriageway_Hazards	70
Urban_or_Rural_Area	70
Did_Police_Officer_Attend_Scene_of_Accident	70
LSOA_of_Accident_Location	70
Vehicle_Reference	70
Vehicle_Type	70
Towing_and_Articulation	70
Vehicle_Manoeuvre	70
Vehicle_Location-Restricted_Lane	70
Junction_Location	70
Skidding_and_Overturning	70
Hit_Object_in_Carriageway	70
Vehicle_Leaving_Carriageway	70
Hit_Object_off_Carriageway	70
1st_Point_of_Impact	70
Was_Vehicle_Left_Hand_Drive?	70
Journey_Purpose_of_Driver	70
Sex_of_Driver	70
Age_of_Driver	70
Age_Band_of_Driver	70
Engine_Capacity_(CC)	70
Propulsion_Code	70
Age_of_Vehicle	70
Driver_IMD_Decile	70
Driver_Home_Area_Type	70
dtype: int64	

Inference: From this we can see that 70 persons from our data have been met with most severe accident which endangered their life.

```
sns.distplot(data['Speed_limit'])
```

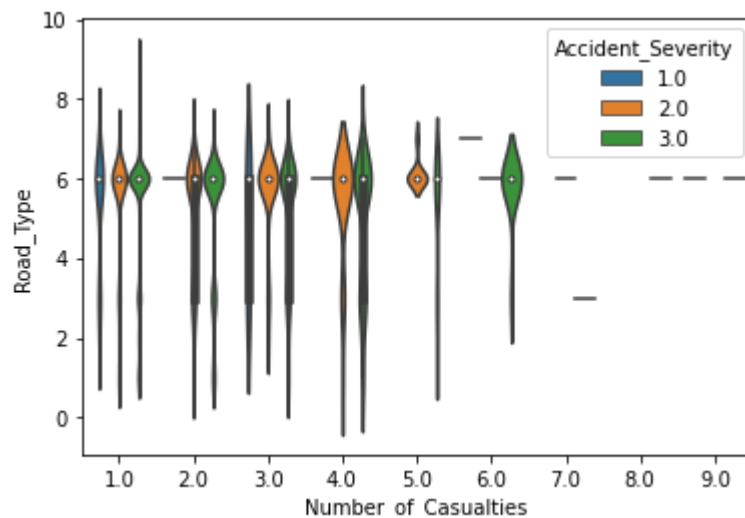
```
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: FutureWarning: `distplot` is deprecated and will be removed in a future version. Use `displot` instead.
warnings.warn(msg, FutureWarning)
<matplotlib.axes._subplots.AxesSubplot at 0x7f94074f4d10>
```



In our dataset, the Speed limit is distributed in the range of 20-70.

```
sns.violinplot(data['Number_of_Casualties'], data['Road_Type'], hue=data['Accident_Severity'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword arguments: {'x': 'Number_of_Casualties', 'y': 'Road_Type', 'hue': 'Accident_Severity'}.
FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7f9400ad2750>
```



## Identifying dependent and independent features:

```
y=data['Accident_Severity']
x=data[['Longitude','Latitude','Number_of_Vehicles','Number_of_Casualties','Road_Type','Speed_limit','Weather_Conditions','Road_Surface_Condition
```

## Splitting the dataset into train and test:

```
from sklearn.model_selection import train_test_split

x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=101)
```

## Use Ensemble learning VotingClassifier technique:

```
#Use ensemble learning:
from sklearn.ensemble import VotingClassifier
from sklearn.linear_model import LogisticRegression
model1=LogisticRegression(max_iter=100)
from sklearn.tree import DecisionTreeClassifier
model2=DecisionTreeClassifier(random_state=101)
from sklearn import svm
#Create a svm Classifier
model3 = svm.SVC(kernel='linear') # Linear Kernel
model=VotingClassifier(estimators=[('log',model1),('dt',model2),('svm',model3)],voting='hard')
model.fit(x_train,y_train)
pred=model.predict(x_test)
```

/usr/local/lib/python3.7/dist-packages/sklearn/linear\_model/\_logistic.py:818: ConvergenceWarning: STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max\_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

[https://scikit-learn.org/stable/modules/linear\\_model.html#logistic-regression](https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

extra\_warning\_msg=\_LOGISTIC\_SOLVER\_CONVERGENCE\_MSG,



```
import pandas as pd
pred=pd.DataFrame(pred)
pred
```

	0
0	3.0
1	3.0
2	3.0
3	3.0
4	3.0

Evaluating our model:

**2801** 3.0

```
from sklearn.metrics import accuracy_score
accuracy_score(y_test, pred)
```

0.8970064148253742

**2804** 3.0

Inference: We get accuracy to be 90% approximately => our model works very well.

2806 rows x 1 columns

Saving the trained model:

```
import pickle
```

```
filename="accident_severity_model.pkl"
pickle.dump(model, open(filename, 'wb'))
```

Loading the model:

```
loaded_model=pickle.load(open('accident_severity_model.pkl', 'rb'))
```

loaded\_model

```
VotingClassifier(estimators=[('log', LogisticRegression()),
                             ('dt', DecisionTreeClassifier(random_state=101)),
                             ('svm', SVC(kernel='linear'))])
```

