

PRODIGY_DS_05

```
[1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
```

```
[2]: df=pd.read_csv("/content/RTA Dataset.csv")
df.head()
```

```
[2]:
```

	Time	Day_of_week	Age_band_of_driver	Sex_of_driver	Educational_level	\
0	17:02:00	Monday	18-30	Male	Above high school	
1	17:02:00	Monday	31-50	Male	Junior high school	
2	17:02:00	Monday	18-30	Male	Junior high school	
3	1:06:00	Sunday	18-30	Male	Junior high school	
4	1:06:00	Sunday	18-30	Male	Junior high school	

	Vehicle_driver_relation	Driving_experience	Type_of_vehicle	\
0	Employee	1-2yr	Automobile	
1	Employee	Above 10yr	Public (> 45 seats)	
2	Employee	1-2yr	Lorry (41?100Q)	
3	Employee	5-10yr	Public (> 45 seats)	
4	Employee	2-5yr	NaN	

	Owner_of_vehicle	Service_year_of_vehicle	... Vehicle_movement	\
0	Owner	Above 10yr	... Going straight	
1	Owner	5-10yrs	... Going straight	
2	Owner	NaN	... Going straight	
3	Governmental	NaN	... Going straight	
4	Owner	5-10yrs	... Going straight	

	Casualty_class	Sex_of_casualty	Age_band_of_casualty	Casualty_severity	\
0	na	na	na	na	
1	na	na	na	na	
2	Driver or rider	Male	31-50	3	
3	Pedestrian	Female	18-30	3	
4	na	na	na	na	

	Work_of_casualty	Fitness_of_casualty	Pedestrian_movement \
0	NaN	NaN	Not a Pedestrian
1	NaN	NaN	Not a Pedestrian
2	Driver	NaN	Not a Pedestrian
3	Driver	Normal	Not a Pedestrian
4	NaN	NaN	Not a Pedestrian

	Cause_of_accident	Accident_severity
0	Moving Backward	Slight Injury
1	Overtaking	Slight Injury
2	Changing lane to the left	Serious Injury
3	Changing lane to the right	Slight Injury
4	Overtaking	Slight Injury

[5 rows x 32 columns]

```
[3]: df.shape
```

```
[3]: (5993, 32)
```

```
[4]: df.describe()
```

```
[4]:
```

	Number_of_vehicles_involved	Number_of_casualties
count	5992.000000	5992.000000
mean	1.972964	1.465621
std	0.624651	0.928860
min	1.000000	1.000000
25%	2.000000	1.000000
50%	2.000000	1.000000
75%	2.000000	2.000000
max	6.000000	8.000000

```
[5]: df.describe(include="all")
```

```
[5]:
```

	Time	Day_of_week	Age_band_of_driver	Sex_of_driver \
count	5993	5993	5993	5992
unique	927	7	6	3
top	16:00:00	Friday	31-50	Male
freq	57	975	2042	5483
mean	NaN	NaN	NaN	NaN
std	NaN	NaN	NaN	NaN
min	NaN	NaN	NaN	NaN
25%	NaN	NaN	NaN	NaN
50%	NaN	NaN	NaN	NaN
75%	NaN	NaN	NaN	NaN
max	NaN	NaN	NaN	NaN

	Educational_level	Vehicle_driver_relation	Driving_experience	\
count	5643	5746	5595	
unique	7	4	7	
top	Junior high school	Employee	5-10yr	
freq	3696	4630	1665	
mean	NaN	NaN	NaN	
std	NaN	NaN	NaN	
min	NaN	NaN	NaN	
25%	NaN	NaN	NaN	
50%	NaN	NaN	NaN	
75%	NaN	NaN	NaN	
max	NaN	NaN	NaN	

	Type_of_vehicle	Owner_of_vehicle	Service_year_of_vehicle	...	\
count	5511	5762	4019	...	
unique	17	4	6	...	
top	Automobile	Owner	Unknown	...	
freq	1573	5088	1377	...	
mean	NaN	NaN	NaN	...	
std	NaN	NaN	NaN	...	
min	NaN	NaN	NaN	...	
25%	NaN	NaN	NaN	...	
50%	NaN	NaN	NaN	...	
75%	NaN	NaN	NaN	...	
max	NaN	NaN	NaN	...	

	Vehicle_movement	Casualty_class	Sex_of_casualty	Age_band_of_casualty	\
count	5870	5992	5992	5992	
unique	13	4	3	6	
top	Going straight	Driver or rider	Male	na	
freq	4033	2344	2507	2105	
mean	NaN	NaN	NaN	NaN	
std	NaN	NaN	NaN	NaN	
min	NaN	NaN	NaN	NaN	
25%	NaN	NaN	NaN	NaN	
50%	NaN	NaN	NaN	NaN	
75%	NaN	NaN	NaN	NaN	
max	NaN	NaN	NaN	NaN	

	Casualty_severity	Work_of_casualty	Fitness_of_casualty	\
count	5992	4430	4692	
unique	4	7	5	
top	3	Driver	Normal	
freq	3395	2862	4656	
mean	NaN	NaN	NaN	
std	NaN	NaN	NaN	

min	NaN	NaN	NaN
25%	NaN	NaN	NaN
50%	NaN	NaN	NaN
75%	NaN	NaN	NaN
max	NaN	NaN	NaN

	Pedestrian_movement	Cause_of_accident	Accident_severity
count	5992	5992	5992
unique	9	20	3
top	Not a Pedestrian	No distancing	Slight Injury
freq	5523	1104	5177
mean	NaN	NaN	NaN
std	NaN	NaN	NaN
min	NaN	NaN	NaN
25%	NaN	NaN	NaN
50%	NaN	NaN	NaN
75%	NaN	NaN	NaN
max	NaN	NaN	NaN

[11 rows x 32 columns]

[6] : df.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 5993 entries, 0 to 5992

Data columns (total 32 columns):

#	Column	Non-Null Count	Dtype
0	Time	5993 non-null	object
1	Day_of_week	5993 non-null	object
2	Age_band_of_driver	5993 non-null	object
3	Sex_of_driver	5992 non-null	object
4	Educational_level	5643 non-null	object
5	Vehicle_driver_relation	5746 non-null	object
6	Driving_experience	5595 non-null	object
7	Type_of_vehicle	5511 non-null	object
8	Owner_of_vehicle	5762 non-null	object
9	Service_year_of_vehicle	4019 non-null	object
10	Defect_of_vehicle	3791 non-null	object
11	Area_accident_occured	5874 non-null	object
12	Lanes_or_Medians	5793 non-null	object
13	Road_allignment	5925 non-null	object
14	Types_of_Junction	5992 non-null	object
15	Road_surface_type	5911 non-null	object
16	Road_surface_conditions	5992 non-null	object
17	Light_conditions	5992 non-null	object
18	Weather_conditions	5992 non-null	object

19	Type_of_collision	5927	non-null	object
20	Number_of_vehicles_involved	5992	non-null	float64
21	Number_of_casualties	5992	non-null	float64
22	Vehicle_movement	5870	non-null	object
23	Casualty_class	5992	non-null	object
24	Sex_of_casualty	5992	non-null	object
25	Age_band_of_casualty	5992	non-null	object
26	Casualty_severity	5992	non-null	object
27	Work_of_casualty	4430	non-null	object
28	Fitness_of_casualty	4692	non-null	object
29	Pedestrian_movement	5992	non-null	object
30	Cause_of_accident	5992	non-null	object
31	Accident_severity	5992	non-null	object

dtypes: float64(2), object(30)
memory usage: 1.5+ MB

```
[7]: df.duplicated().sum()
```

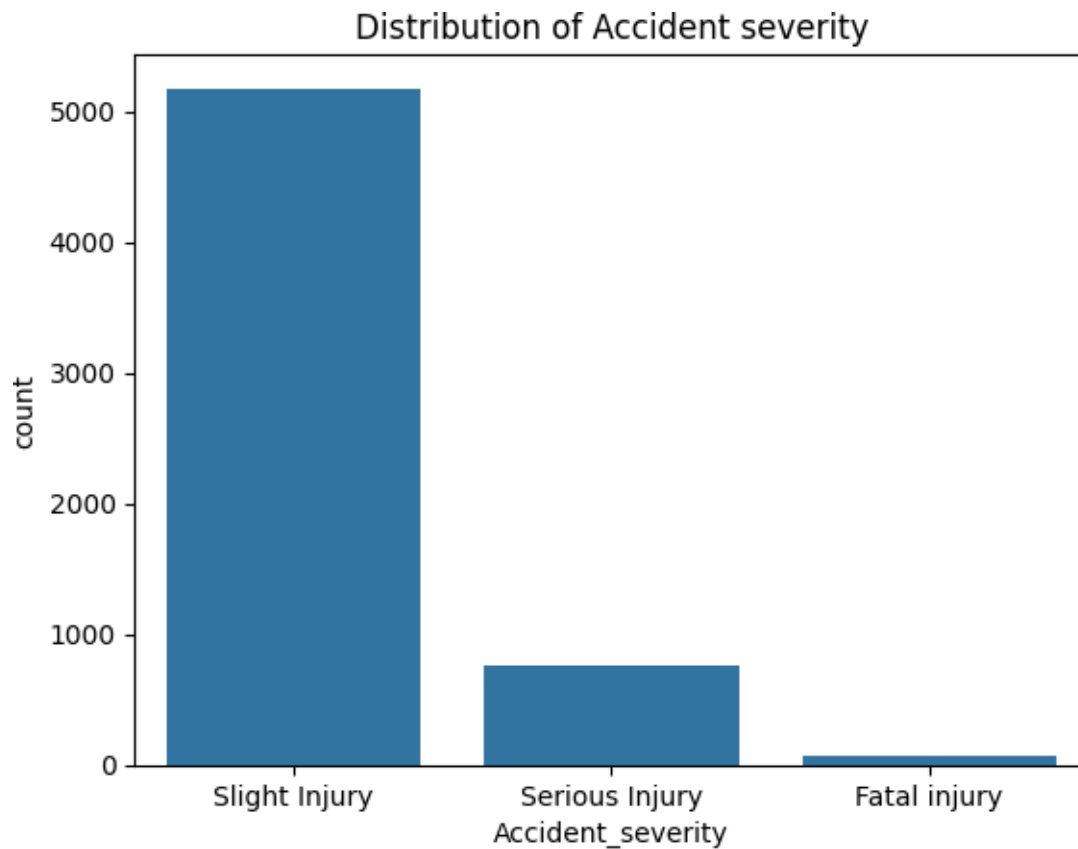
```
[7]: 0
```

```
[8]: df['Accident_severity'].value_counts()
```

```
[8]: Accident_severity
Slight Injury      5177
Serious Injury     753
Fatal injury       62
Name: count, dtype: int64
```

```
[9]: sns.countplot(x = df['Accident_severity'])
plt.title('Distribution of Accident severity')
```

```
[9]: Text(0.5, 1.0, 'Distribution of Accident severity')
```



```
[10]: df.isna().sum()
```

```
[10]: Time                0
      Day_of_week         0
      Age_band_of_driver  0
      Sex_of_driver       1
      Educational_level    350
      Vehicle_driver_relation 247
      Driving_experience    398
      Type_of_vehicle      482
      Owner_of_vehicle     231
      Service_year_of_vehicle 1974
      Defect_of_vehicle    2202
      Area_accident_occured 119
      Lanes_or_Medians     200
      Road_allignment      68
      Types_of_Junction    1
      Road_surface_type     82
      Road_surface_conditions 1
      Light_conditions      1
```

```

Weather_conditions      1
Type_of_collision      66
Number_of_vehicles_involved  1
Number_of_casualties    1
Vehicle_movement      123
Casualty_class          1
Sex_of_casualty         1
Age_band_of_casualty    1
Casualty_severity       1
Work_of_casualty       1563
Fitness_of_casualty     1301
Pedestrian_movement     1
Cause_of_accident       1
Accident_severity       1
dtype: int64

```

```

[11] : df.drop(['Service_year_of_vehicle','Defect_of_vehicle','Work_of_casualty',
, 'Fitness_of_casualty','Time'],
axis = 1, inplace = True)
df.head()

```

```

[11] :   Day_of_week  Age_band_of_driver  Sex_of_driver  Educational_level \
0      Monday      18-30      Male  Above high school
1      Monday      31-50      Male  Junior high school
2      Monday      18-30      Male  Junior high school
3      Sunday      18-30      Male  Junior high school
4      Sunday      18-30      Male  Junior high school

```

```

   Vehicle_driver_relation  Driving_experience  Type_of_vehicle \
0      Employee      1-2yr      Automobile
1      Employee  Above 10yr  Public (> 45 seats)
2      Employee      1-2yr  Lorry (41?100Q)
3      Employee      5-10yr  Public (> 45 seats)
4      Employee      2-5yr      NaN

```

```

   Owner_of_vehicle  Area_accident_occured  Lanes_or_Medians  ... \
0      Owner      Residential areas      NaN  ...
1      Owner      Office areas  Undivided Two way  ...
2      Owner      Recreational areas      other  ...
3  Governmental      Office areas      other  ...
4      Owner      Industrial areas      other  ...

```

```

   Number_of_vehicles_involved  Number_of_casualties  Vehicle_movement \
0      2.0      2.0  Going straight
1      2.0      2.0  Going straight
2      2.0      2.0  Going straight
3      2.0      2.0  Going straight

```

4		2.0	2.0	Going straight
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	Casualty_class	Sex_of_casualty	Age_band_of_casualty	Casualty_severity	\
0	na	na	na	na	
1	na	na	na	na	
2	Driver or rider	Male	31-50	3	
3	Pedestrian	Female	18-30	3	
4	na	na	na	na	

	Pedestrian_movement	Cause_of_accident	Accident_severity
0	Not a Pedestrian	Moving Backward	Slight Injury
1	Not a Pedestrian	Overtaking	Slight Injury
2	Not a Pedestrian	Changing lane to the left	Serious Injury
3	Not a Pedestrian	Changing lane to the right	Slight Injury
4	Not a Pedestrian	Overtaking	Slight Injury

[5 rows x 27 columns]

```
[12] : categorical=[i for i in df.columns if df[i].dtype=='O']
print('The categorical variables are',categorical)
```

The categorical variables are ['Day_of_week', 'Age_band_of_driver', 'Sex_of_driver', 'Educational_level', 'Vehicle_driver_relation', 'Driving_experience', 'Type_of_vehicle', 'Owner_of_vehicle', 'Area_accident_occured', 'Lanes_or_Medians', 'Road_alignment', 'Types_of_Junction', 'Road_surface_type', 'Road_surface_conditions', 'Light_conditions', 'Weather_conditions', 'Type_of_collision', 'Vehicle_movement', 'Casualty_class', 'Sex_of_casualty', 'Age_band_of_casualty', 'Casualty_severity', 'Pedestrian_movement', 'Cause_of_accident', 'Accident_severity']

```
[13] : for i in categorical:
        df[i].fillna(df[i].mode()[0],inplace=True)
```

```
[14] : df.isna().sum()
```

```
[14]: Day_of_week          0
      Age_band_of_driver  0
      Sex_of_driver      0
      Educational_level   0
      Vehicle_driver_relation  0
      Driving_experience   0
      Type_of_vehicle     0
      Owner_of_vehicle    0
      Area_accident_occured  0
      Lanes_or_Medians    0
      Road_alignment      0
```



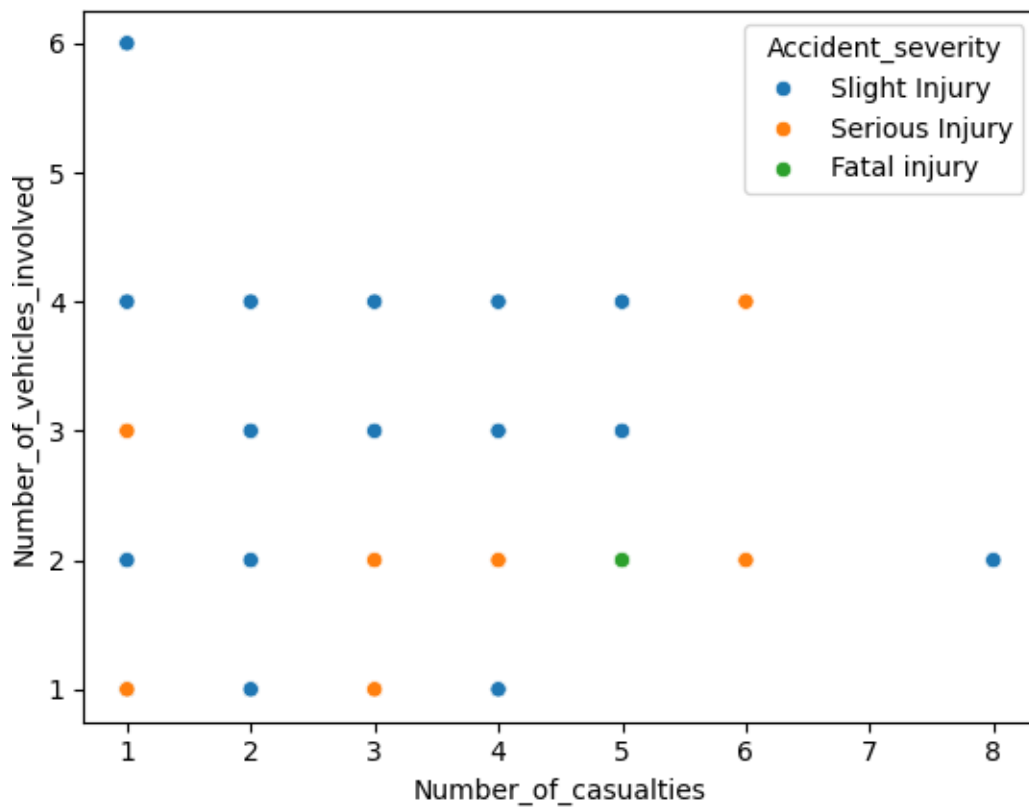
```

Types_of_Junction          0
Road_surface_type          0
Road_surface_conditions    0
Light_conditions           0
Weather_conditions         0
Type_of_collision          0
Number_of_vehicles_involved 1
Number_of_casualties       1
Vehicle_movement           0
Casualty_class             0
Sex_of_casualty            0
Age_band_of_casualty       0
Casualty_severity          0
Pedestrian_movement        0
Cause_of_accident          0
Accident_severity          0
dtype: int64

```

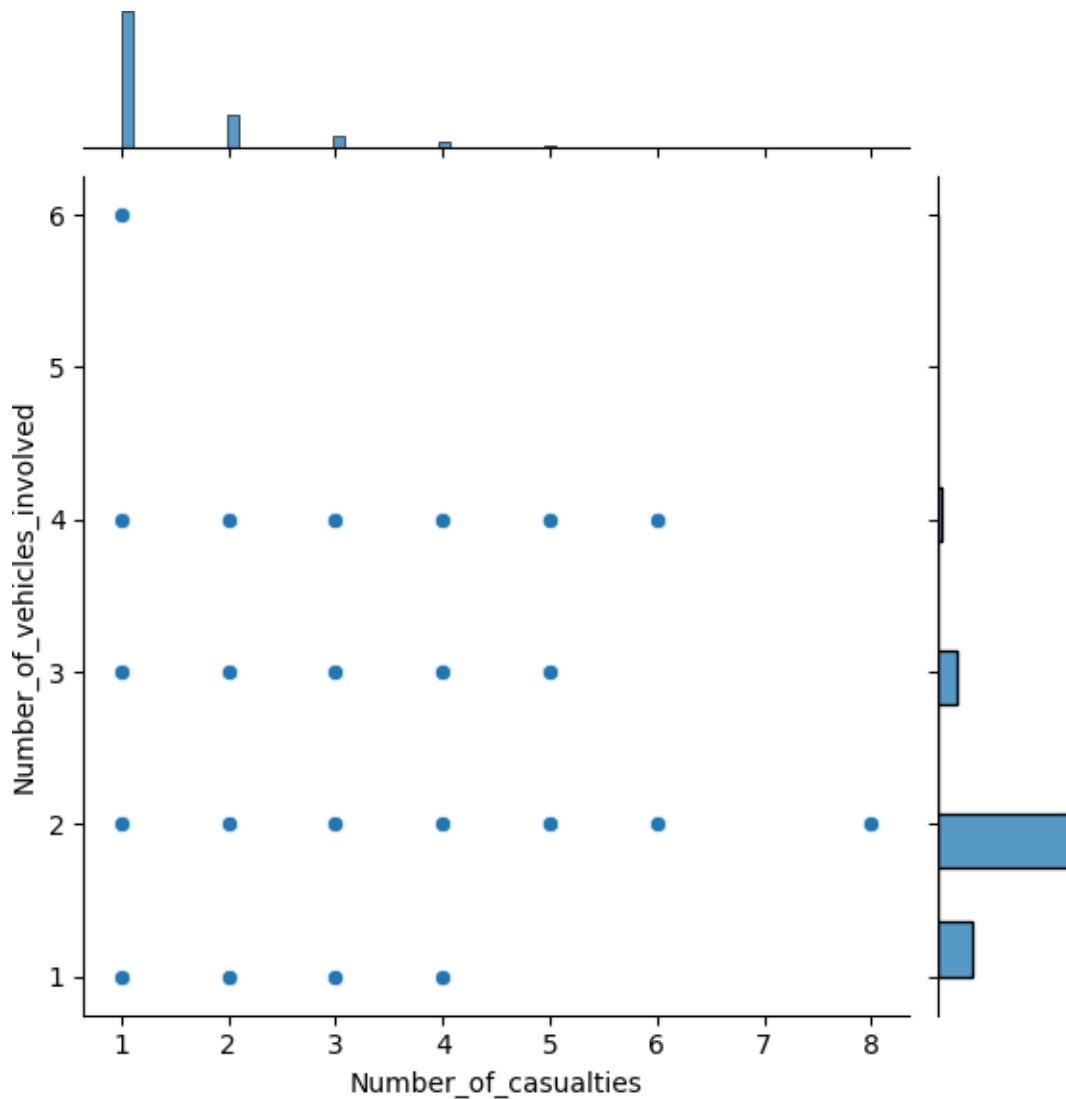
```
[15]: sns.scatterplot(x=df['Number_of_casualties'],
, y=df['Number_of_vehicles_involved'], hue=df['Accident_severity'])
```

```
[15]: <Axes: xlabel='Number_of_casualties', ylabel='Number_of_vehicles_involved'>
```



```
[16]: sns.jointplot(x='Number_of_casualties',y='Number_of_vehicles_involved',data=df)
```

```
[16]: <seaborn.axisgrid.JointGrid at 0x799c62415790>
```

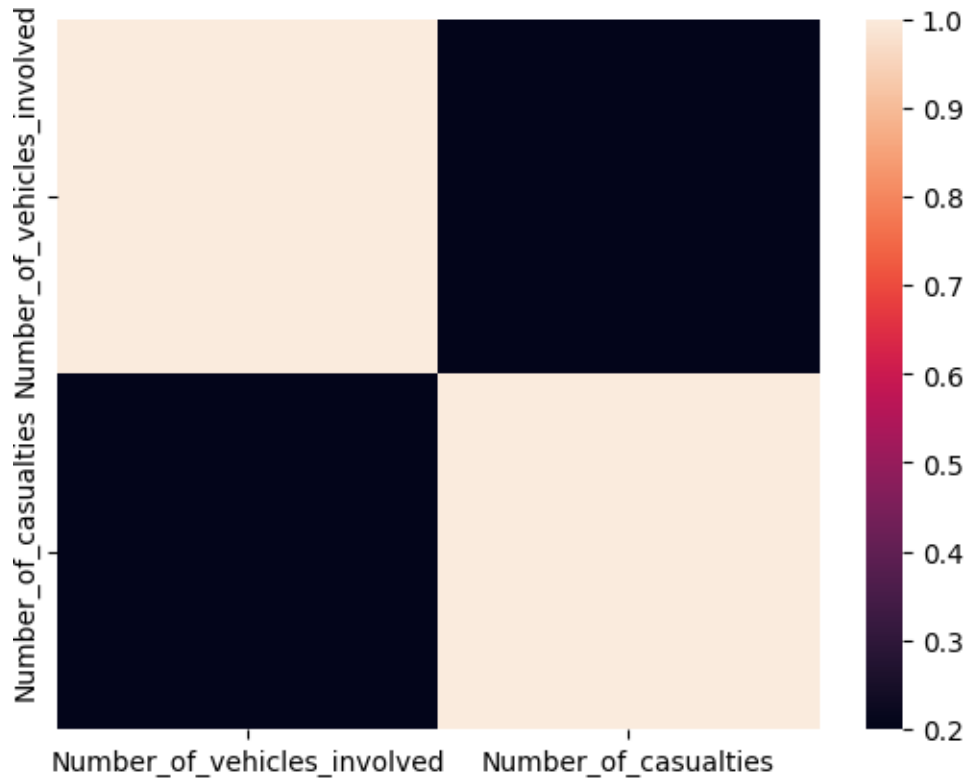


```
[18]: numerical_df = df.select_dtypes(include=np.number) # Select numerical columns
correlation_matrix = numerical_df.corr()
print(correlation_matrix)
```

	Number_of_vehicles_involved	Number_of_casualties
Number_of_vehicles_involved	1.000000	0.199775
Number_of_casualties	0.199775	1.000000

```
[20]: #plotting the correlation using heatmap
sns.heatmap(df.select_dtypes(include=np.number).corr())
```

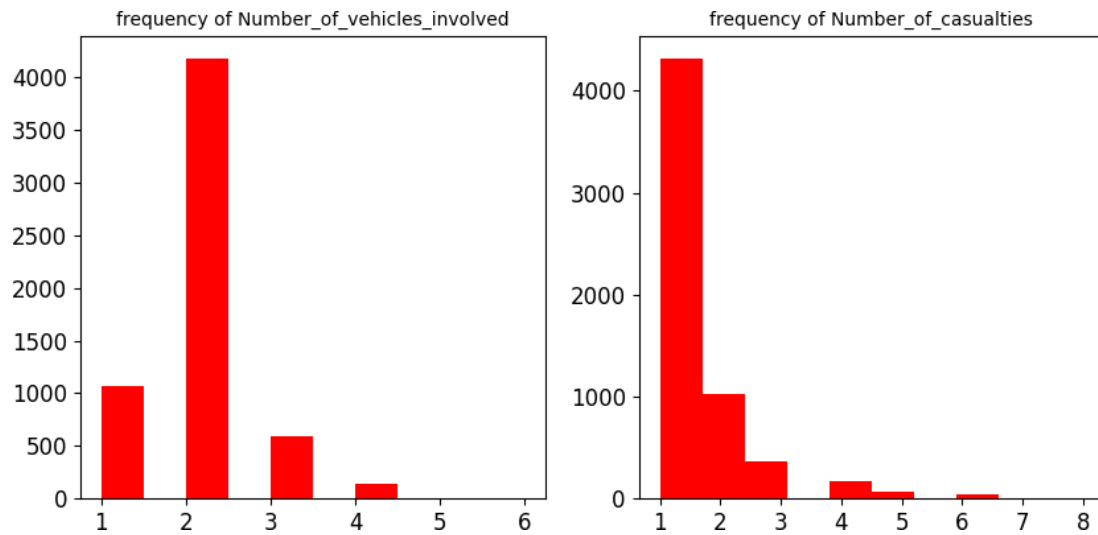
[20] : <Axes: >



```
[21] : numerical=[i for i in df.columns if df[i].dtype!='O']
print('The numerica variables are',numerical)
```

The numerica variables are ['Number_of_vehicles_involved', 'Number_of_casualties']

```
[22] : plt.figure(figsize=(10,10))
plotnumber = 1
for i in numerical:
    if plotnumber <= df.shape[1]:
        ax1 = plt.subplot(2,2,plotnumber)
        plt.hist(df[i],color='red')
        plt.xticks(fontsize=12)
        plt.yticks(fontsize=12)
        plt.title('frequency of '+i, fontsize=10)
    plotnumber +=1
```



```
[23] : #count plot for categorical values
plt.figure(figsize=(10,200))
plotnumber = 1

for col in categorical:
    if plotnumber <= df.shape[1] and col!='Pedestrian_movement':
        ax1 = plt.subplot(28,1,plotnumber)
        sns.countplot(data=df, y=col, palette='muted')
        plt.xticks(fontsize=12)
        plt.yticks(fontsize=12)
        plt.title(col.title(), fontsize=14)
        plt.xlabel("")
        plt.ylabel("")
        plotnumber += 1
```



```
[24] : df.dtypes
```

```
[24] : Day_of_week           object
      Age_band_of_driver  object
      Sex_of_driver       object
      Educational_level    object
      Vehicle_driver_relation object
      Driving_experience   object
      Type_of_vehicle      object
      Owner_of_vehicle     object
      Area_accident_occured object
      Lanes_or_Medians     object
      Road_alignment       object
      Types_of_Junction    object
      Road_surface_type    object
      Road_surface_conditions object
      Light_conditions     object
      Weather_conditions   object
      Type_of_collision     object
      Number_of_vehicles_involved float64
      Number_of_casualties float64
      Vehicle_movement      object
      Casualty_class        object
      Sex_of_casualty       object
      Age_band_of_casualty  object
      Casualty_severity     object
      Pedestrian_movement  object
      Cause_of_accident     object
      Accident_severity     object
      dtype: object
```

```
[25] : #importing label encoing module
      from sklearn.preprocessing import LabelEncoder
      le=LabelEncoder()

      #creating a new data frame from performing the chi2 analysis
      df1=pd.DataFrame()

      #adding all the categorical columns except the output to new data frame
      for i in categorical:
          if i!= 'Accident_severity':
              df1[i]=le.fit_transform(df[i])
```

```
[26] : #confirming the data type
      df1.info()
```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 5993 entries, 0 to 5992

Data columns (total 24 columns):

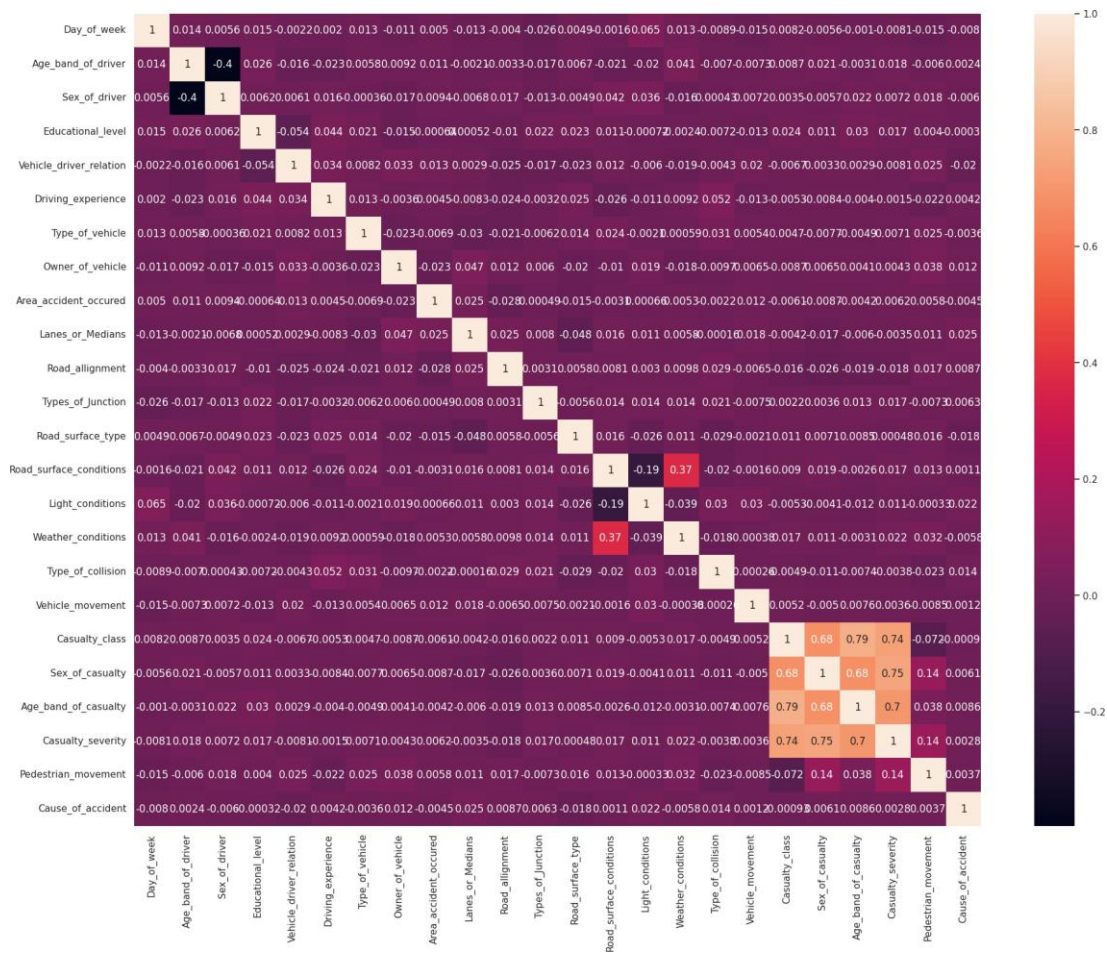
#	Column	Non-Null Count		Dtype
0	Day_of_week	5993	non-null	int64
1	Age_band_of_driver	5993	non-null	int64
2	Sex_of_driver	5993	non-null	int64
3	Educational_level	5993	non-null	int64
4	Vehicle_driver_relation	5993	non-null	int64
5	Driving_experience	5993	non-null	int64
6	Type_of_vehicle	5993	non-null	int64
7	Owner_of_vehicle	5993	non-null	int64
8	Area_accident_occured	5993	non-null	int64
9	Lanes_or_Medians	5993	non-null	int64
10	Road_allignment	5993	non-null	int64
11	Types_of_Junction	5993	non-null	int64
12	Road_surface_type	5993	non-null	int64
13	Road_surface_conditions	5993	non-null	int64
14	Light_conditions	5993	non-null	int64
15	Weather_conditions	5993	non-null	int64
16	Type_of_collision	5993	non-null	int64
17	Vehicle_movement	5993	non-null	int64
18	Casualty_class	5993	non-null	int64
19	Sex_of_casualty	5993	non-null	int64
20	Age_band_of_casualty	5993	non-null	int64
21	Casualty_severity	5993	non-null	int64
22	Pedestrian_movement	5993	non-null	int64
23	Cause_of_accident	5993	non-null	int64

dtypes: int64(24)

memory usage: 1.1 MB

```
[27]: plt.figure(figsize=(22,17))
      sns.set(font_scale=1)
      sns.heatmap(df1.corr(), annot=True)
```

[27]: <Axes: >



[28]: *#label encoded data set*
df1.head()

```
[28]: Day_of_week Age_band_of_driver Sex_of_driver Educational_level \
0 1 0 1 0
1 1 1 1 4
2 1 0 1 4
3 3 0 1 4
4 3 0 1 4

Vehicle_driver_relation Driving_experience Type_of_vehicle \
0 0 0 0
1 0 3 11
2 0 0 5
3 0 2 11
4 0 1 0
```


	Owner_of_vehicle	Area_accident_occured	Lanes_or_Medians	...	\
0	3	9	2	...	
1	3	6	4	...	
2	3	1	6	...	
3	0	6	6	...	
4	3	4	6	...	

	Light_conditions	Weather_conditions	Type_of_collision	Vehicle_movement	\
0	3	2	3	2	
1	3	2	8	2	
2	3	2	2	2	
3	0	2	8	2	
4	0	2	8	2	

	Casualty_class	Sex_of_casualty	Age_band_of_casualty	Casualty_severity	\
0	3	2	5	3	
1	3	2	5	3	
2	0	1	1	2	
3	2	0	0	2	
4	3	2	5	3	

	Pedestrian_movement	Cause_of_accident
0	5	9
1	5	16
2	5	0
3	5	1
4	5	16

[5 rows x 24 columns]

```
[29]: #import chi2 test
from sklearn.feature_selection import chi2
f_p_values=chi2(df1,df['Accident_severity'])
```

```
[30]: #f_p_values will return Fscore and pvalues
f_p_values
```

```
[30]: (array([ 1.80883262, 9.45043594, 0.01922965, 0.07231767, 13.96771244,
7.19300683, 0.32360606, 0.73984515, 0.81039617, 3.47759726,
0.01541708, 7.6266644 , 3.36859125, 4.06800158, 4.31967441,
1.49649648, 7.48554952, 9.05708919, 0.04410499, 0.51213415,
7.99699866, 0.15700447, 0.22789597, 2.87309999]),
array([4.04778081e-01, 8.86878059e-03, 9.90431251e-01, 9.64487088e-01,
9.26722657e-04, 2.74194294e-02, 8.50608734e-01, 6.90787813e-01,
6.66844704e-01, 1.75731392e-01, 9.92321095e-01, 2.20744996e-02,
1.85575100e-01, 1.30811125e-01, 1.15343897e-01, 4.73194750e-01,
2.36882826e-02, 1.07963778e-02, 9.78188886e-01, 7.74090044e-01,
```

1.83431452e-02, 9.24499995e-01, 8.92304370e-01, 2.37746573e-01]))

```
[31] : #for better understanding and ease of access adding them to a new dataframe
f_p_values1=pd.DataFrame({'features':df1.columns, 'Fscore': f_p_values[0],
, 'Pvalues':f_p_values[1]})
f_p_values1
```

```
[31]:
```

	features	Fscore	Pvalues
0	Day_of_week	1.808833	0.404778
1	Age_band_of_driver	9.450436	0.008869
2	Sex_of_driver	0.019230	0.990431
3	Educational_level	0.072318	0.964487
4	Vehicle_driver_relation	13.967712	0.000927
5	Driving_experience	7.193007	0.027419
6	Type_of_vehicle	0.323606	0.850609
7	Owner_of_vehicle	0.739845	0.690788
8	Area_accident_occured	0.810396	0.666845
9	Lanes_or_Medians	3.477597	0.175731
10	Road_allignment	0.015417	0.992321
11	Types_of_Junction	7.626664	0.022074
12	Road_surface_type	3.368591	0.185575
13	Road_surface_conditions	4.068002	0.130811
14	Light_conditions	4.319674	0.115344
15	Weather_conditions	1.496496	0.473195
16	Type_of_collision	7.485550	0.023688
17	Vehicle_movement	9.057089	0.010796
18	Casualty_class	0.044105	0.978189
19	Sex_of_casualty	0.512134	0.774090
20	Age_band_of_casualty	7.996999	0.018343
21	Casualty_severity	0.157004	0.924500
22	Pedestrian_movement	0.227896	0.892304
23	Cause_of_accident	2.873100	0.237747

```
[32] : #since we want lower Pvalues we are sorting the features
f_p_values1.sort_values(by='Pvalues',ascending=True)
```

```
[32]:
```

	features	Fscore	Pvalues
4	Vehicle_driver_relation	13.967712	0.000927
1	Age_band_of_driver	9.450436	0.008869
17	Vehicle_movement	9.057089	0.010796
20	Age_band_of_casualty	7.996999	0.018343
11	Types_of_Junction	7.626664	0.022074
16	Type_of_collision	7.485550	0.023688
5	Driving_experience	7.193007	0.027419
14	Light_conditions	4.319674	0.115344
13	Road_surface_conditions	4.068002	0.130811
9	Lanes_or_Medians	3.477597	0.175731

12	Road_surface_type	3.368591	0.185575
23	Cause_of_accident	2.873100	0.237747
0	Day_of_week	1.808833	0.404778
15	Weather_conditions	1.496496	0.473195
8	Area_accident_occured	0.810396	0.666845
7	Owner_of_vehicle	0.739845	0.690788
19	Sex_of_casualty	0.512134	0.774090
6	Type_of_vehicle	0.323606	0.850609
22	Pedestrian_movement	0.227896	0.892304
21	Casualty_severity	0.157004	0.924500
3	Educational_level	0.072318	0.964487
18	Casualty_class	0.044105	0.978189
2	Sex_of_driver	0.019230	0.990431
10	Road_allignment	0.015417	0.992321

```
[33] : #after evaluating we are removing lesser important columns and storing to a new
data frame
df2=df.drop(['Owner_of_vehicle', 'Type_of_vehicle', 'Road_surface_conditions',
'Pedestrian_movement',
'Casualty_severity','Educational_level','Day_of_week','Sex_of_driver','Road_allignment',
'Sex_of_casualty'],axis=1)
df2.head()
```

```
[33]: Age_band_of_driver Vehicle_driver_relation Driving_experience \
0      18-30      Employee      1-2yr
1      31-50      Employee      Above 10yr
2      18-30      Employee      1-2yr
3      18-30      Employee      5-10yr
4      18-30      Employee      2-5yr

Area_accident_occured      Lanes_or_Medians \
0      Residential areas      Two-way (divided with broken lines road marking)
1      Office areas      Undivided Two way
2      Recreational areas      other
3      Office areas      other
4      Industrial areas      other

Types_of_Junction Road_surface_type      Light_conditions \
0      No junction      Asphalt roads      Daylight
1      No junction      Asphalt roads      Daylight
2      No junction      Asphalt roads      Daylight
3      Y Shape      Earth roads      Darkness - lights lit
4      Y Shape      Asphalt roads      Darkness - lights lit

Weather_conditions      Type_of_collision \
0      Normal      Collision with roadside-parked vehicles
```

1	Normal	Vehicle with vehicle collision
2	Normal	Collision with roadside objects
3	Normal	Vehicle with vehicle collision
4	Normal	Vehicle with vehicle collision

	Number_of_vehicles_involved	Number_of_casualties	Vehicle_movement \
0	2.0	2.0	Going straight
1	2.0	2.0	Going straight
2	2.0	2.0	Going straight
3	2.0	2.0	Going straight
4	2.0	2.0	Going straight

	Casualty_class	Age_band_of_casualty	Cause_of_accident \
0	na	na	Moving Backward
1	na	na	Overtaking
2	Driver or rider	31-50	Changing lane to the left
3	Pedestrian	18-30	Changing lane to the right
4	na	na	Overtaking

	Accident_severity
0	Slight Injury
1	Slight Injury
2	Serious Injury
3	Slight Injury
4	Slight Injury

```
[34] : df2.shape
```

```
[34]: (5993, 17)
```

```
[35] : df2.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 5993 entries, 0 to 5992
```

```
Data columns (total 17 columns):
```

#	Column	Non-Null Count	Dtype
0	Age_band_of_driver	5993 non-null	object
1	Vehicle_driver_relation	5993 non-null	object
2	Driving_experience	5993 non-null	object
3	Area_accident_occured	5993 non-null	object
4	Lanes_or_Medians	5993 non-null	object
5	Types_of_Junction	5993 non-null	object
6	Road_surface_type	5993 non-null	object
7	Light_conditions	5993 non-null	object
8	Weather_conditions	5993 non-null	object
9	Type_of_collision	5993 non-null	object

10	Number_of_vehicles_involved	5992	non-null	float64
11	Number_of_casualties	5992	non-null	float64
12	Vehicle_movement	5993	non-null	object
13	Casualty_class	5993	non-null	object
14	Age_band_of_casualty	5993	non-null	object
15	Cause_of_accident	5993	non-null	object
16	Accident_severity	5993	non-null	object

dtypes: float64(2), object(15)
memory usage: 796.1+ KB

```
[36] : #to check distinct values in each categorical columns we are storing them to a new variable
categorical_new=[i for i in df2.columns if df2[i].dtype=='O']
print(categorical_new)
```

```
['Age_band_of_driver', 'Vehicle_driver_relation', 'Driving_experience',
'Area_accident_occured', 'Lanes_or_Medians', 'Types_of_Junction',
'Road_surface_type', 'Light_conditions', 'Weather_conditions',
'Type_of_collision', 'Vehicle_movement', 'Casualty_class',
'Age_band_of_casualty', 'Cause_of_accident', 'Accident_severity']
```

```
[37] : for i in categorical_new:
        print(df2[i].value_counts())
```

```
Age_band_of_driver
31-50      2042
18-30      1974
Unknown     942
Over 51      695
Under 18     339
Under 1        1
Name: count, dtype: int64
Vehicle_driver_relation
Employee   4877
Owner     1042
Other       60
Unknown     14
Name: count, dtype: int64
Driving_experience
5-10yr     2063
2-5yr     1254
Above 10yr  1111
1-2yr       841
Below 1yr   646
No Licence   60
unknown     18
Name: count, dtype: int64
Area_accident_occured
```

Other	2012
Office areas	1674
Residential areas	964
Church areas	510
Industrial areas	223
School areas	203
Recreational areas	164
Outside rural areas	119
Hospital areas	59
Market areas	29
Rural village areas	18
Unknown	9
Rural village areasOffice areas	8
Recreational areas	1
Name: count, dtype: int64	
Lanes_or_Medians	
Two-way (divided with broken lines road marking)	2356
Undivided Two way	1844
other	788
Double carriageway (median)	507
One way	411
Two-way (divided with solid lines road marking)	62
Unknown	25
Name: count, dtype: int64	
Types_of_Junction	
Y Shape	2400
No junction	1966
Crossing	1178
Other	215
Unknown	115
O Shape	89
T Shape	30
Name: count, dtype: int64	
Road_surface_type	
Asphalt roads	5587
Earth roads	169
Gravel roads	117
Other	80
Asphalt roads with some distress	40
Name: count, dtype: int64	
Light_conditions	
Daylight	4206
Darkness – lights lit	1693
Darkness – no lighting	83
Darkness – lights unlit	11
Name: count, dtype: int64	
Weather_conditions	
Normal	4949

Raining	578
Unknown	190
Other	139
Cloudy	48
Windy	43
Snow	21
Raining and Windy	20
Fog or mist	5
Name: count, dtype: int64	
Type_of_collision	
Vehicle with vehicle collision	4319
Collision with roadside objects	871
Collision with pedestrians	440
Rollover	189
Collision with animals	95
Collision with roadside-parked vehicles	30
Fall from vehicles	17
Unknown	14
Other	13
With Train	5
Name: count, dtype: int64	
Vehicle_movement	
Going straight	4156
Moving Backward	475
Other	389
Reversing	281
Turnover	243
Getting off	152
Entering a junction	86
Unknown	81
Overtaking	44
Stopping	29
Waiting to go	25
U-Turn	22
Parked	10
Name: count, dtype: int64	
Casualty_class	
Driver or rider	2345
na	2105
Pedestrian	838
Passenger	705
Name: count, dtype: int64	
Age_band_of_casualty	
na	2106
18-30	1423
31-50	1204
Under 18	618
Over 51	567

```

5          75
Name: count, dtype: int64
Cause_of_accident
No distancing          1105
Changing lane to the right    898
Changing lane to the left    704
Driving carelessly          672
No priority to vehicle      579
Moving Backward           553
No priority to pedestrian   365
Other                     235
Overtaking                208
Driving under the influence of drugs  151
Driving to the left        135
Getting off the vehicle improperly    92
Driving at high speed        87
Overturning                71
Turnover                   40
Overloading                30
Overspeed                  28
Drunk driving              14
Unknown                    13
Improper parking           13
Name: count, dtype: int64
Accident_severity
Slight Injury    5178
Serious Injury   753
Fatal injury     62
Name: count, dtype: int64

```

```

[39]: #get_dummies
dummy=pd.get_dummies(df2[['Age_band_of_driver', 'Vehicle_driver_relation',
    'Driving_experience',
    'Area_accident_occured', 'Lanes_or_Medians',
    'Types_of_Junction', 'Road_surface_type',
    'Light_conditions', 'Weather_conditions',
    'Type_of_collision', 'Vehicle_movement',
    'Casualty_class', 'Age_band_of_casualty',
    'Cause_of_accident']],drop_first=True)
dummy.head()

```

```

[39] :   Age_band_of_driver_31-50   Age_band_of_driver_Over 51 \
0                False                False
1                 True                False
2                False                False
3                False                False
4                False                False

```


	Age_band_of_driver_Under 18	Age_band_of_driver_Under 18 \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

	Age_band_of_driver_Unknown	Vehicle_driver_relation_Other \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

	Vehicle_driver_relation_Owner	Vehicle_driver_relation_Unknown \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

	Driving_experience_2-5yr	Driving_experience_5-10yr	... \
0	False	False	...
1	False	False	...
2	False	False	...
3	False	True	...
4	True	False	...

	Cause_of_accident_No distancing \
0	False
1	False
2	False
3	False
4	False

	Cause_of_accident_No priority to pedestrian \
0	False
1	False
2	False
3	False
4	False

	Cause_of_accident_No priority to vehicle	Cause_of_accident_Other \
0	False	False
1	False	False
2	False	False

3		False	False
4		False	False

	Cause_of_accident_Overloading	Cause_of_accident_Overspeed	\
0	False	False	
1	False	False	
2	False	False	
3	False	False	
4	False	False	

	Cause_of_accident_Overtaking	Cause_of_accident_Overturning	\
0	False	False	
1	True	False	
2	False	False	
3	False	False	
4	True	False	

	Cause_of_accident_Turnover	Cause_of_accident_Unknown
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

[5 rows x 102 columns]

```
[40] : #concatinate dummy and old data frame
df3=pd.concat([df2,dummy],axis=1)
df3.head()
```

[40] :	Age_band_of_driver	Vehicle_driver_relation	Driving_experience	\
0	18-30	Employee	1-2yr	
1	31-50	Employee	Above 10yr	
2	18-30	Employee	1-2yr	
3	18-30	Employee	5-10yr	
4	18-30	Employee	2-5yr	

	Area_accident_occured	Lanes_or_Medians	\
0	Residential areas	Two-way (divided with broken lines road marking)	
1	Office areas	Undivided Two way	
2	Recreational areas	other	
3	Office areas	other	
4	Industrial areas	other	

	Types_of_Junction	Road_surface_type	Light_conditions	\
0	No junction	Asphalt roads	Daylight	
1	No junction	Asphalt roads	Daylight	

2	No junction	Asphalt roads	Daylight
3	Y Shape	Earth roads	Darkness – lights lit
4	Y Shape	Asphalt roads	Darkness – lights lit

	Weather_conditions	Type_of_collision	...	\
0	Normal	Collision with roadside-parked vehicles	...	
1	Normal	Vehicle with vehicle collision	...	
2	Normal	Collision with roadside objects	...	
3	Normal	Vehicle with vehicle collision	...	
4	Normal	Vehicle with vehicle collision	...	

	Cause_of_accident_No distancing	\
0	False	
1	False	
2	False	
3	False	
4	False	

	Cause_of_accident_No priority to pedestrian	\
0	False	
1	False	
2	False	
3	False	
4	False	

	Cause_of_accident_No priority to vehicle	Cause_of_accident_Other	\
0	False	False	
1	False	False	
2	False	False	
3	False	False	
4	False	False	

	Cause_of_accident_Overloading	Cause_of_accident_Overspeed	\
0	False	False	
1	False	False	
2	False	False	
3	False	False	
4	False	False	

	Cause_of_accident_Overtaking	Cause_of_accident_Overturning	\
0	False	False	
1	True	False	
2	False	False	
3	False	False	
4	True	False	

	Cause_of_accident_Turnover	Cause_of_accident_Unknown
--	----------------------------	---------------------------

0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

[5 rows x 119 columns]

[41] : *#dropping dummied columns*

```
df3.drop(['Age_band_of_driver', 'Vehicle_driver_relation',
        'Driving_experience', 'Area_accident_occured', 'Lanes_or_Medians',
        'Types_of_Junction', 'Road_surface_type', 'Light_conditions',
        'Weather_conditions', 'Type_of_collision',
        'Vehicle_movement', 'Casualty_class', 'Age_band_of_casualty',
        'Cause_of_accident'],axis=1,inplace=True)
df3.head()
```

[41]:

	Number_of_vehicles_involved	Number_of_casualties	Accident_severity \
0	2.0	2.0	Slight Injury
1	2.0	2.0	Slight Injury
2	2.0	2.0	Serious Injury
3	2.0	2.0	Slight Injury
4	2.0	2.0	Slight Injury

	Age_band_of_driver_31-50	Age_band_of_driver_Over 51 \
0	False	False
1	True	False
2	False	False
3	False	False
4	False	False

	Age_band_of_driver_Under 1	Age_band_of_driver_Under 18 \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

	Age_band_of_driver_Unknown	Vehicle_driver_relation_Other \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

	Vehicle_driver_relation_Owner	... Cause_of_accident_No distancing \
0	False	False

1	False	...	False
2	False	...	False
3	False	...	False
4	False	...	False

Cause_of_accident_No priority to pedestrian \		
0		False
1		False
2		False
3		False
4		False

Cause_of_accident_No priority to vehicle		Cause_of_accident_Other \	
0	False		False
1	False		False
2	False		False
3	False		False
4	False		False

Cause_of_accident_Overloading		Cause_of_accident_Overspeed \	
0	False		False
1	False		False
2	False		False
3	False		False
4	False		False

Cause_of_accident_Overtaking		Cause_of_accident_Overturning \	
0	False		False
1	True		False
2	False		False
3	False		False
4	True		False

Cause_of_accident_Turnover		Cause_of_accident_Unknown	
0	False		False
1	False		False
2	False		False
3	False		False
4	False		False

[5 rows x 105 columns]

```
[42]: x=df3.drop(['Accident_severity'],axis=1)
      x.shape
```

[42]: (5993, 104)

```
[43] : x.head()
```

```
[43]:
```

	Number_of_vehicles_involved	Number_of_casualties	\
0	2.0	2.0	
1	2.0	2.0	
2	2.0	2.0	
3	2.0	2.0	
4	2.0	2.0	

	Age_band_of_driver_31-50	Age_band_of_driver_Over 51	\
0	False	False	
1	True	False	
2	False	False	
3	False	False	
4	False	False	

	Age_band_of_driver_Under 1	Age_band_of_driver_Under 18	\
0	False	False	
1	False	False	
2	False	False	
3	False	False	
4	False	False	

	Age_band_of_driver_Unknown	Vehicle_driver_relation_Other	\
0	False	False	
1	False	False	
2	False	False	
3	False	False	
4	False	False	

	Vehicle_driver_relation_Owner	Vehicle_driver_relation_Unknown	...	\
0	False	False	...	
1	False	False	...	
2	False	False	...	
3	False	False	...	
4	False	False	...	

	Cause_of_accident_No distancing	\
0	False	
1	False	
2	False	
3	False	
4	False	

	Cause_of_accident_No priority to pedestrian	\
0	False	
1	False	

2	False
3	False
4	False

	Cause_of_accident_No priority to vehicle	Cause_of_accident_Other \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

	Cause_of_accident_Overloading	Cause_of_accident_Overspeed \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

	Cause_of_accident_Overtaking	Cause_of_accident_Overturning \
0	False	False
1	True	False
2	False	False
3	False	False
4	True	False

	Cause_of_accident_Turnover	Cause_of_accident_Unknown
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

[5 rows x 104 columns]

```
[44]: y=df3.iloc[:,2]
      y.head()
```

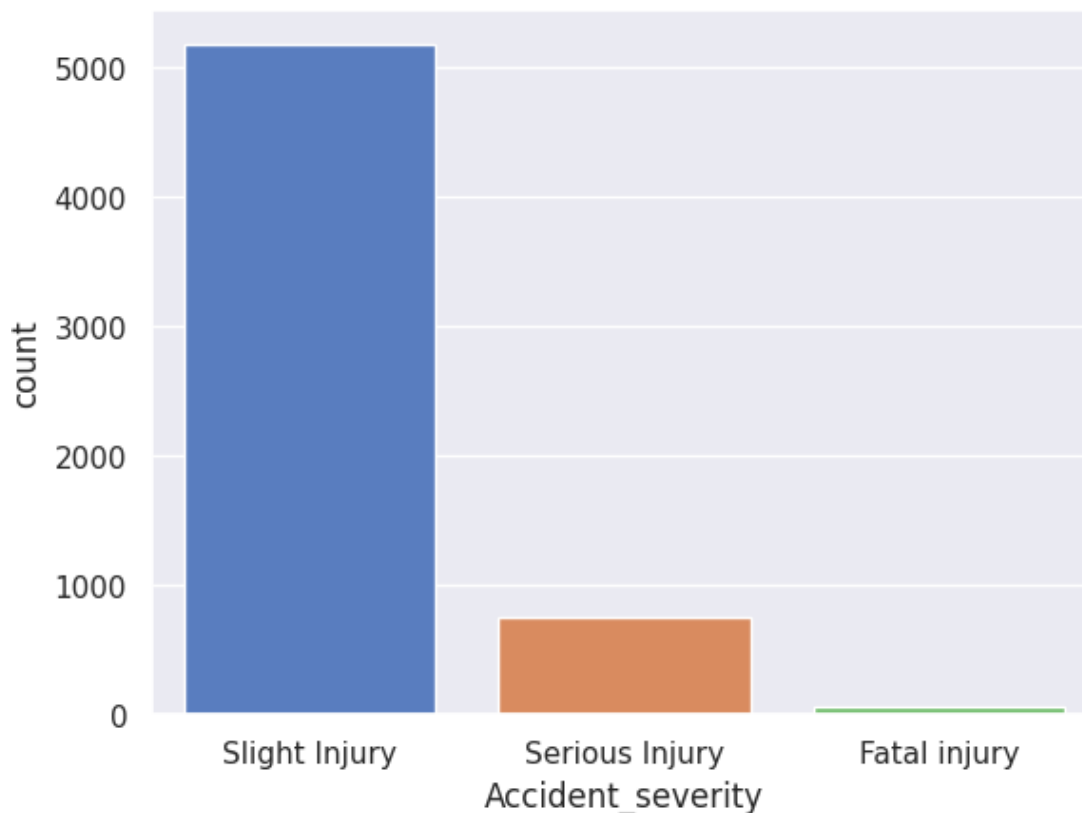
```
[44]: 0    Slight Injury
      1    Slight Injury
      2    Serious Injury
      3    Slight Injury
      4    Slight Injury
      Name: Accident_severity, dtype: object
```

```
[45]: #checking the count of each item in the output column
      y.value_counts()
```

```
[45]: Accident_severity
      Slight Injury      5178
      Serious Injury    753
      Fatal injury      62
      Name: count, dtype: int64
```

```
[46]: #plotting count plot using seaborn
      sns.countplot(x = y, palette='muted')
```

```
[46]: <Axes: xlabel='Accident_severity', ylabel='count'>
```



```
[49]: # Impute missing values using SimpleImputer
      from sklearn.impute import SimpleImputer

      # Create an imputer object with your desired strategy (e.g., mean, median,
      # most_frequent)
      imputer = SimpleImputer(strategy='most_frequent') # Replace with your
      # preferred strategy

      # Fit the imputer on your data and transform it
      x_imputed = imputer.fit_transform(x)
```



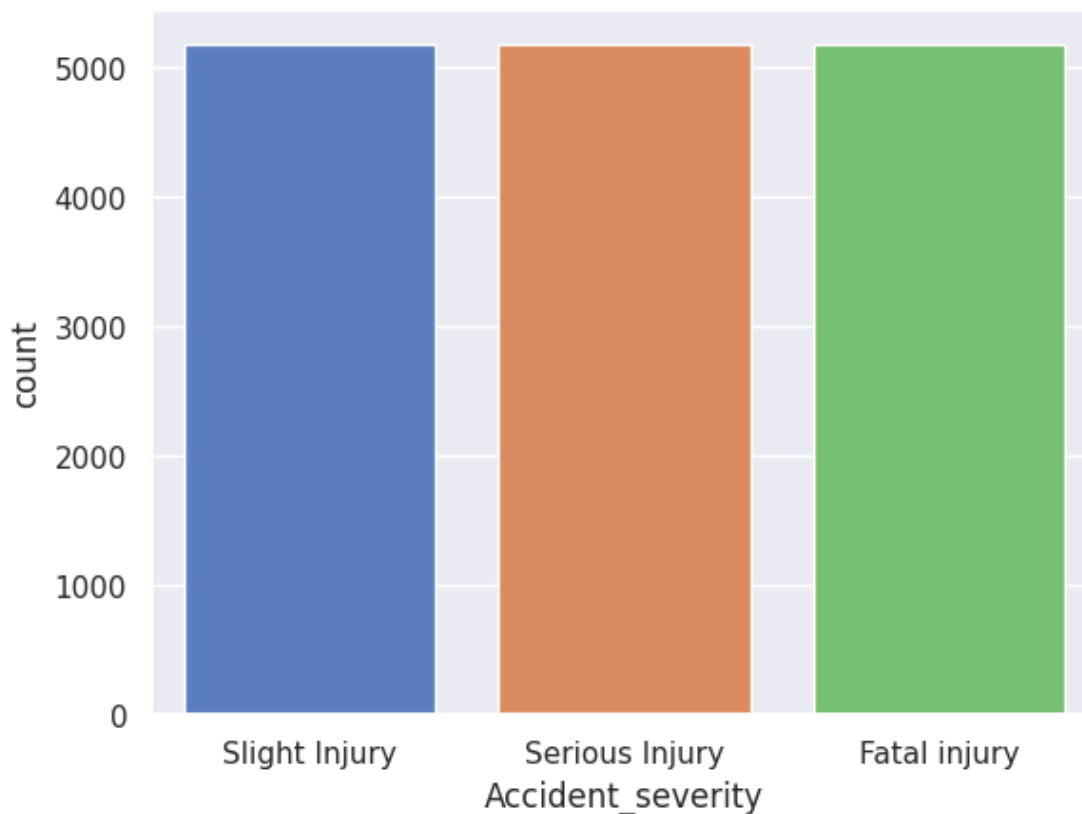
```
# Now, apply SMOTE on the imputed data
xo, yo = oversample.fit_resample(x_imputed, y)
```

```
[50]: #checking the oversampling output
y1=pd.DataFrame(yo)
y1.value_counts()
```

```
[50]: Accident_severity
Fatal injury      5178
Serious Injury    5178
Slight Injury     5178
Name: count, dtype: int64
```

```
[51]: sns.countplot(x = yo, palette='muted')
```

```
[51] : <Axes: xlabel='Accident_severity', ylabel='count'>
```



```
[52] : #converting data to training data and testing data
from sklearn.model_selection import train_test_split
#splitting 70% of the data to training data and 30% of data to testing data
```

```
x_train,x_test,y_train,y_test=train_test_split(xo,yo,test_size=0.30,random_state=42)
```

```
[53] : print(x_train.shape,x_test.shape,y_train.shape,y_test.shape)
```

```
(10873, 104) (4661, 104) (10873,) (4661,)
```

```
[54] : #KNN model alg
from sklearn.neighbors import KNeighborsClassifier
model_KNN=KNeighborsClassifier(n_neighbors=5)
model_KNN.fit(x_train,y_train)
```

```
[54] : KNeighborsClassifier()
```

```
[55] : y_pred=model_KNN.predict(x_test)
```

```
[56] : y_pred
```

```
[56] : array(['Serious Injury', 'Serious Injury', 'Serious Injury', ...,
        'Serious Injury', 'Fatal injury', 'Fatal injury'], dtype=object)
```

```
[57] : from sklearn.metrics import_
        classification_report,confusion_matrix,accuracy_score,ConfusionMatrixDisplay
```

```
[58] : report_KNN=classification_report(y_test,y_pred)
        print(report_KNN)
```

	precision	recall	f1-score	support
Fatal injury	0.88	1.00	0.94	1548
Serious Injury	0.67	0.99	0.80	1551
Slight Injury	0.99	0.39	0.56	1562
accuracy			0.79	4661
macro avg	0.85	0.79	0.77	4661
weighted avg	0.85	0.79	0.77	4661

```
[59] : accuracy_KNN=accuracy_score(y_test,y_pred)
        print(accuracy_KNN)
```

```
0.7936065222055353
```

```
[60] : matrix_KNN=confusion_matrix(y_test,y_pred)
        print(matrix_KNN,'\n')
        print(ConfusionMatrixDisplay.from_predictions(y_test,y_pred))
```

```
[[1548    0    0]
 [    8 1539    4]
 [   201  749  612]]
```

<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay object at 0x799c558e1f50>

