

prodigy-ds-05

September 20, 2024

1 Task 5

Analyze traffic accident data to identify patterns related to road conditions, weather, and time of day. Visualize accident hotspots and contributing factors.

Loading Libraries and Data

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

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

```
[5]:      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]

```
[6]: df.shape
```

```
[6]: (12316, 32)
```

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

```
[7]:
```

	Number_of_vehicles_involved	Number_of_casualties
count	12316.000000	12316.000000
mean	2.040679	1.548149
std	0.688790	1.007179
min	1.000000	1.000000
25%	2.000000	1.000000
50%	2.000000	1.000000
75%	2.000000	2.000000
max	7.000000	8.000000

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

```
[8]:
```

	Time	Day_of_week	Age_band_of_driver	Sex_of_driver	\
count	12316	12316	12316	12316	
unique	1074	7	5	3	
top	15:30:00	Friday	18-30	Male	
freq	120	2041	4271	11437	
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	11575	11737	11487	
unique	7	4	7	
top	Junior high school	Employee	5-10yr	
freq	7619	9627	3363	
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	11366	11834	8388	...	
unique	17	4	6	...	
top	Automobile	Owner	Unknown	...	
freq	3205	10459	2883	...	
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	12008	12316	12316	12316	
unique	13	4	3	6	
top	Going straight	Driver or rider	Male	na	
freq	8158	4944	5253	4443	
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	12316	9118	9681
unique	4	7	5
top	3	Driver	Normal
freq	7076	5903	9608
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	12316	12316	12316
unique	9	20	3
top	Not a Pedestrian	No distancing	Slight Injury
freq	11390	2263	10415
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]

```
[9]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12316 entries, 0 to 12315
Data columns (total 32 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Time                                  12316 non-null  object
1   Day_of_week                           12316 non-null  object
2   Age_band_of_driver                     12316 non-null  object
3   Sex_of_driver                           12316 non-null  object
4   Educational_level                       11575 non-null  object
5   Vehicle_driver_relation                 11737 non-null  object
6   Driving_experience                      11487 non-null  object
7   Type_of_vehicle                        11366 non-null  object
8   Owner_of_vehicle                       11834 non-null  object
9   Service_year_of_vehicle                 8388 non-null   object
10  Defect_of_vehicle                       7889 non-null   object
11  Area_accident_occured                   12077 non-null  object
12  Lanes_or_Medians                        11931 non-null  object
```

```

13 Road_alignment          12174 non-null object
14 Types_of_Junction      11429 non-null object
15 Road_surface_type      12144 non-null object
16 Road_surface_conditions 12316 non-null object
17 Light_conditions       12316 non-null object
18 Weather_conditions     12316 non-null object
19 Type_of_collision      12161 non-null object
20 Number_of_vehicles_involved 12316 non-null int64
21 Number_of_casualties   12316 non-null int64
22 Vehicle_movement       12008 non-null object
23 Casualty_class         12316 non-null object
24 Sex_of_casualty        12316 non-null object
25 Age_band_of_casualty   12316 non-null object
26 Casualty_severity      12316 non-null object
27 Work_of_casualty       9118 non-null object
28 Fitness_of_casualty    9681 non-null object
29 Pedestrian_movement    12316 non-null object
30 Cause_of_accident      12316 non-null object
31 Accident_severity      12316 non-null object
dtypes: int64(2), object(30)
memory usage: 3.0+ MB

```

Exploratory Data Analysis

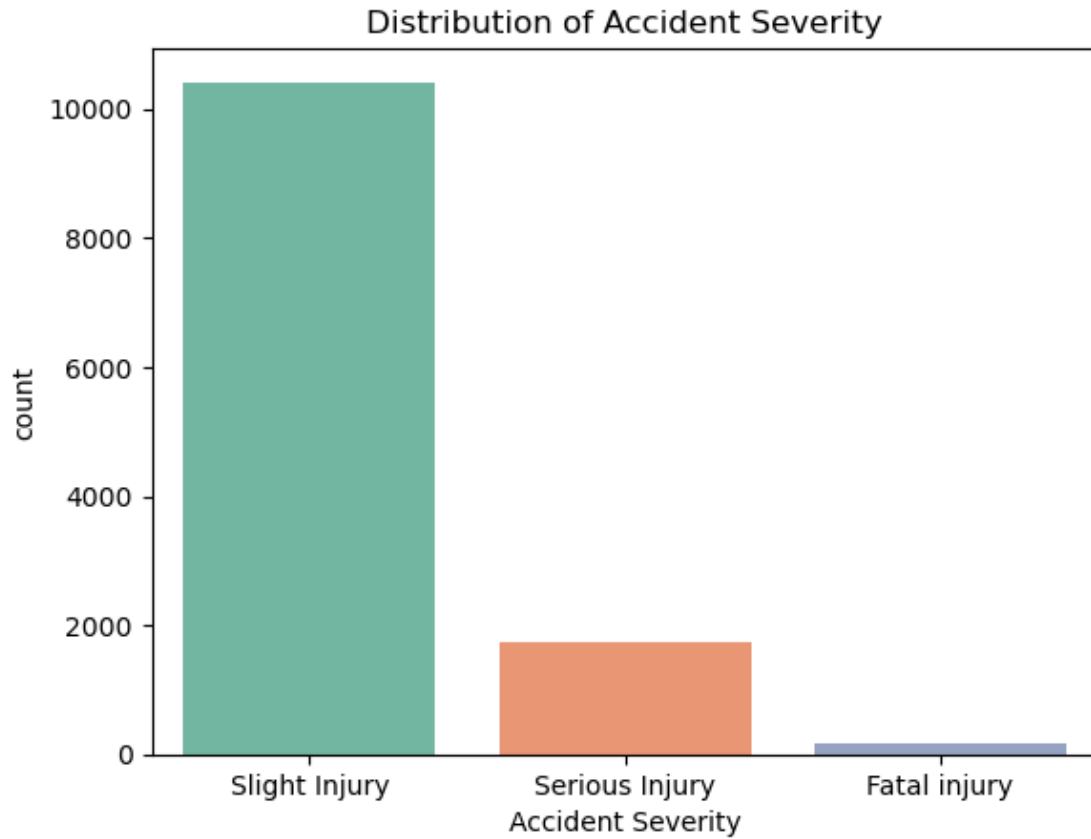
```
[11]: df.duplicated().sum()
```

```
[11]: 0
```

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

```
[12]: Accident_severity
Slight Injury      10415
Serious Injury      1743
Fatal injury        158
Name: count, dtype: int64
```

```
[13]: import seaborn as sns
import matplotlib.pyplot as plt
sns.countplot(x=df['Accident_severity'], palette='Set2')
plt.title('Distribution of Accident Severity')
plt.xlabel('Accident Severity')
plt.show()
```



Handling missing values

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

```
[15]: Time                0
      Day_of_week         0
      Age_band_of_driver  0
      Sex_of_driver       0
      Educational_level    741
      Vehicle_driver_relation 579
      Driving_experience    829
      Type_of_vehicle      950
      Owner_of_vehicle     482
      Service_year_of_vehicle 3928
      Defect_of_vehicle    4427
      Area_accident_occured 239
      Lanes_or_Medians     385
      Road_alignment       142
      Types_of_Junction    887
      Road_surface_type    172
```

```

Road_surface_conditions      0
Light_conditions             0
Weather_conditions           0
Type_of_collision            155
Number_of_vehicles_involved   0
Number_of_casualties         0
Vehicle_movement             308
Casualty_class               0
Sex_of_casualty              0
Age_band_of_casualty         0
Casualty_severity            0
Work_of_casualty             3198
Fitness_of_casualty          2635
Pedestrian_movement          0
Cause_of_accident            0
Accident_severity            0
dtype: int64

```

```

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

```

```

[16]:   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      2      Going straight
1      2      2      Going straight

```

2		2		2	Going straight
3		2		2	Going straight
4		2		2	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	

	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]

```
[17]: 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_allignment', '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']

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

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

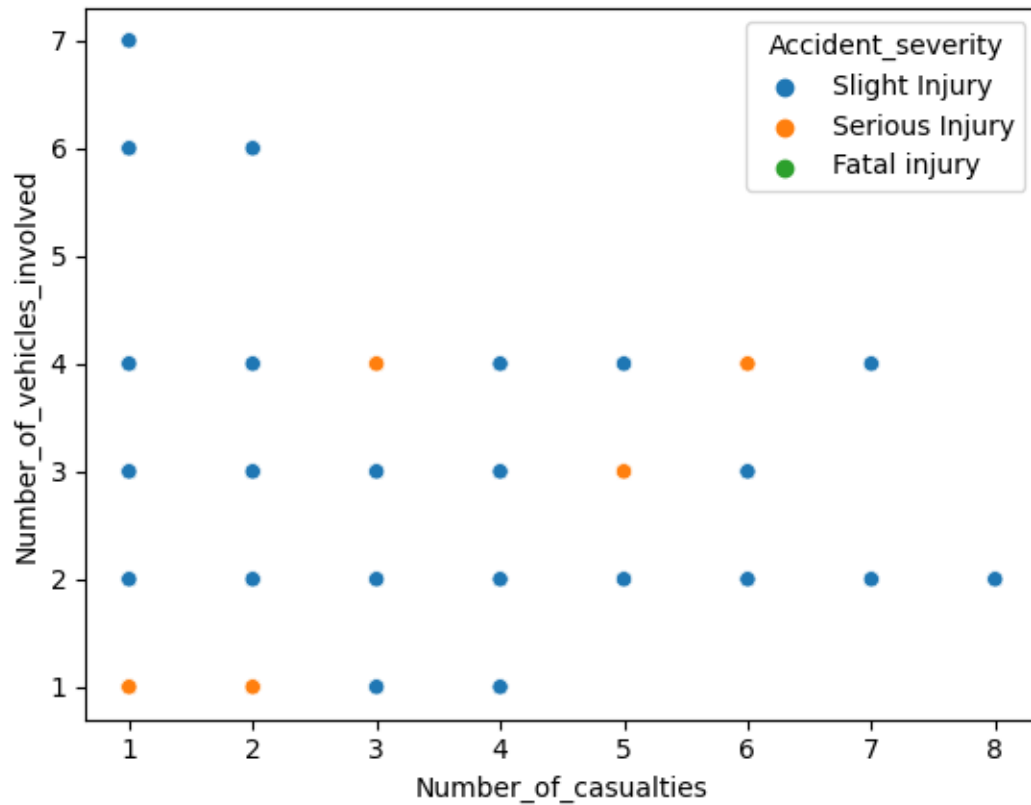
```
[19]: Day_of_week          0
      Age_band_of_driver  0
      Sex_of_driver       0
      Educational_level    0
      Vehicle_driver_relation  0
      Driving_experience    0
      Type_of_vehicle      0
      Owner_of_vehicle     0
      Area_accident_occured 0
```


Lanes_or_Medians	0
Road_allignment	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	0
Number_of_casualties	0
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	

Data Visualization

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

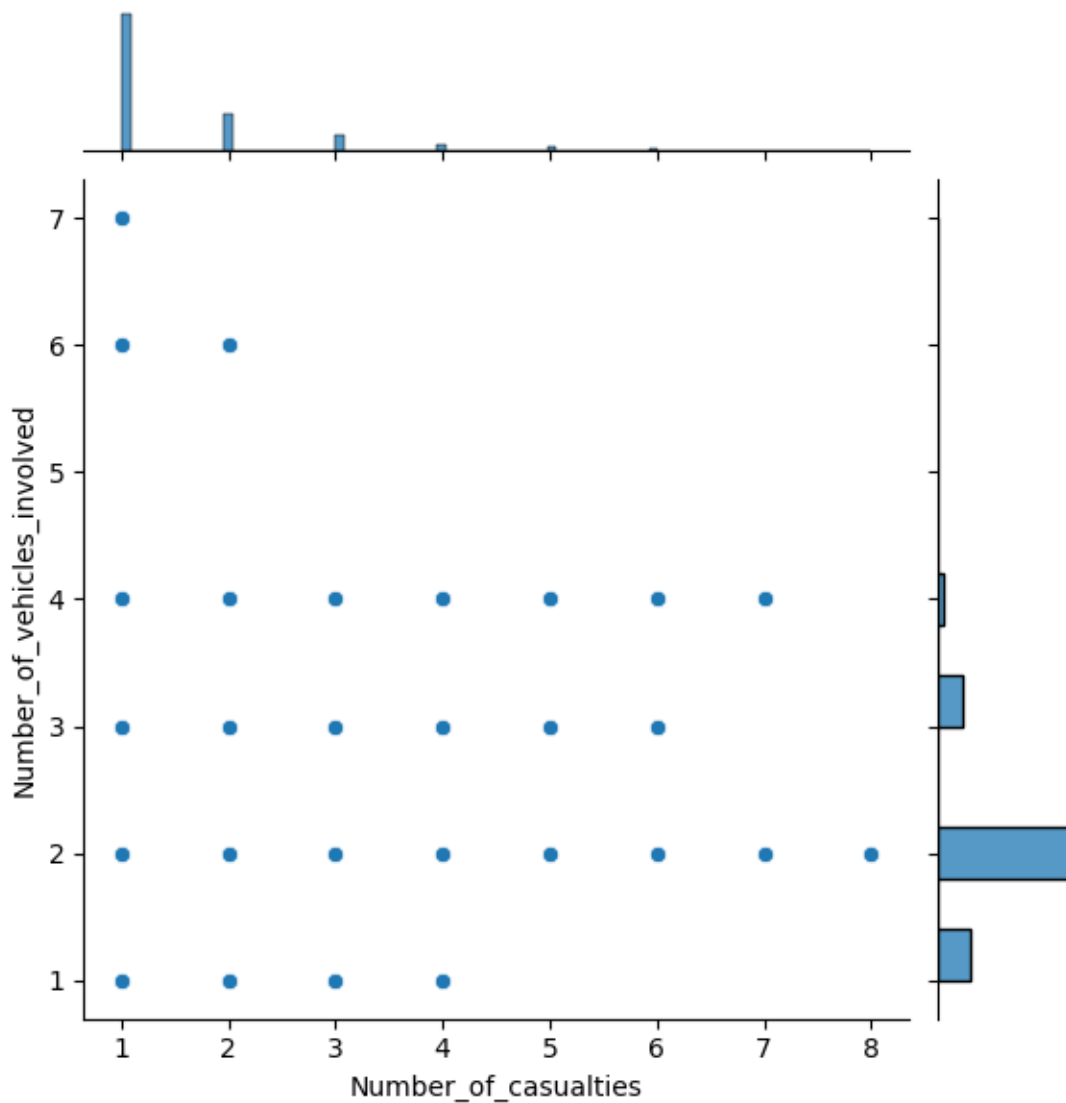
```
[21]: <Axes: xlabel='Number_of_casualties', ylabel='Number_of_vehicles_involved'>
```



There is no visible correlation between Number_of_casualties and Number_of_vehicles_involved columns

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

```
[23]: <seaborn.axisgrid.JointGrid at 0x7fabfff47e10>
```



```
[24]: print(df.info())
numeric_df = df.select_dtypes(include=['number'])
correlation_matrix = numeric_df.corr()
print(correlation_matrix)
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12316 entries, 0 to 12315
Data columns (total 27 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Day_of_week           12316 non-null  object
1   Age_band_of_driver    12316 non-null  object
2   Sex_of_driver         12316 non-null  object
```

3	Educational_level	12316	non-null	object
4	Vehicle_driver_relation	12316	non-null	object
5	Driving_experience	12316	non-null	object
6	Type_of_vehicle	12316	non-null	object
7	Owner_of_vehicle	12316	non-null	object
8	Area_accident_occured	12316	non-null	object
9	Lanes_or_Medians	12316	non-null	object
10	Road_allignment	12316	non-null	object
11	Types_of_Junction	12316	non-null	object
12	Road_surface_type	12316	non-null	object
13	Road_surface_conditions	12316	non-null	object
14	Light_conditions	12316	non-null	object
15	Weather_conditions	12316	non-null	object
16	Type_of_collision	12316	non-null	object
17	Number_of_vehicles_involved	12316	non-null	int64
18	Number_of_casualties	12316	non-null	int64
19	Vehicle_movement	12316	non-null	object
20	Casualty_class	12316	non-null	object
21	Sex_of_casualty	12316	non-null	object
22	Age_band_of_casualty	12316	non-null	object
23	Casualty_severity	12316	non-null	object
24	Pedestrian_movement	12316	non-null	object
25	Cause_of_accident	12316	non-null	object
26	Accident_severity	12316	non-null	object

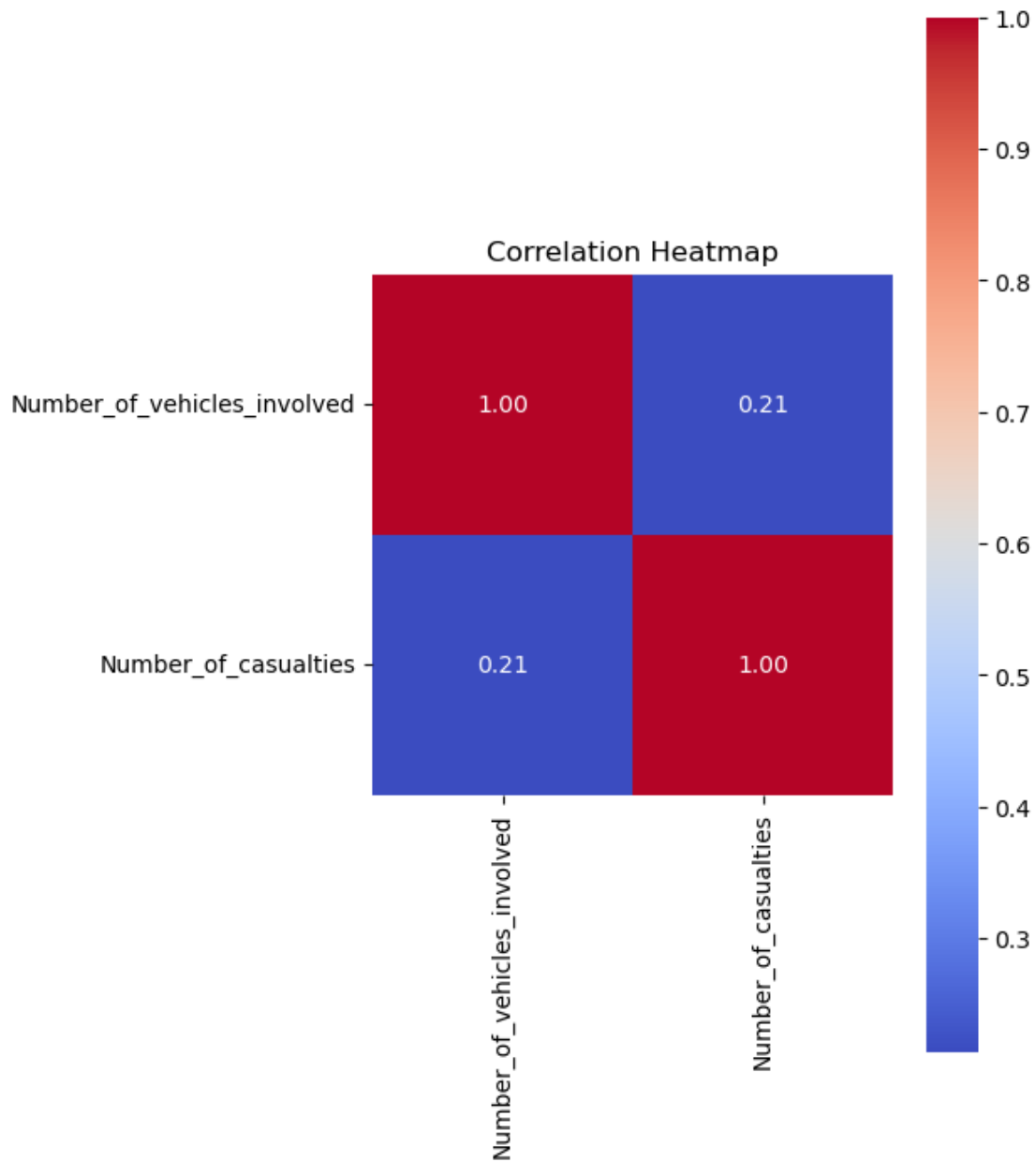
dtypes: int64(2), object(25)

memory usage: 2.5+ MB

None

	Number_of_vehicles_involved	Number_of_casualties
Number_of_vehicles_involved	1.000000	0.213427
Number_of_casualties	0.213427	1.000000

```
[25]: correlation_matrix = df.select_dtypes(include=['number']).corr()
plt.figure(figsize=(5, 8))
sns.heatmap(correlation_matrix, annot=True, fmt=".2f", cmap='coolwarm',
            square=True)
plt.title('Correlation Heatmap')
plt.show()
```



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

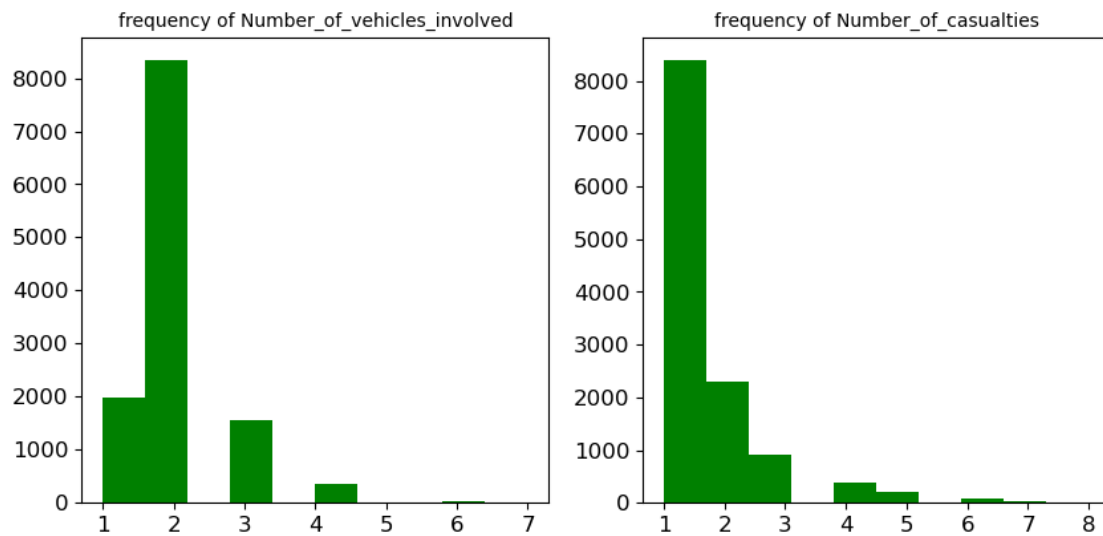
The numerica variables are ['Number_of_vehicles_involved',
'Number_of_casualties']

```
[27]: #distribution for numerical columns  
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='green')
        plt.xticks(fontsize=12)
        plt.yticks(fontsize=12)
        plt.title('frequency of '+i, fontsize=10)
        plotnumber +=1

```



Most accidents are occurred when 2 vehicles are involved and 1 casualty is happened mostly in the accidents.

```

[29]: #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

```


Handling Categorical Values

```
[31]: df.dtypes
```

```
[31]: 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_allignment       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 int64
      Number_of_casualties   int64
      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
```

Since there are so many categorical values, we need to use feature selection. We need to perform label encoding before applying chi square analysis.

```
[33]: from sklearn.preprocessing import LabelEncoder
      le=LabelEncoder()
      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])
```



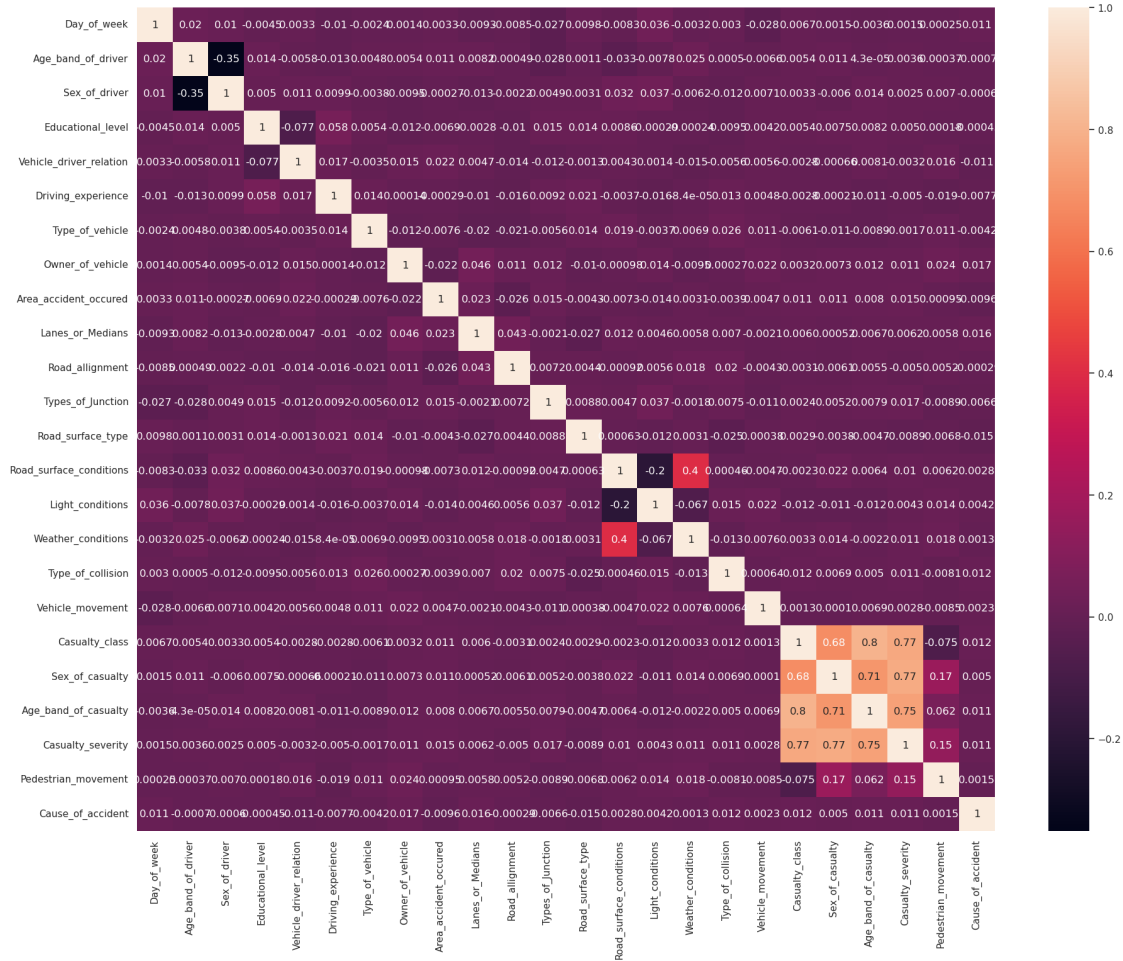
```
[34]: #confirming the data type
df1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12316 entries, 0 to 12315
Data columns (total 24 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Day_of_week                          12316 non-null  int64
1   Age_band_of_driver                  12316 non-null  int64
2   Sex_of_driver                       12316 non-null  int64
3   Educational_level                   12316 non-null  int64
4   Vehicle_driver_relation             12316 non-null  int64
5   Driving_experience                  12316 non-null  int64
6   Type_of_vehicle                    12316 non-null  int64
7   Owner_of_vehicle                   12316 non-null  int64
8   Area_accident_occured              12316 non-null  int64
9   Lanes_or_Medians                   12316 non-null  int64
10  Road_alignment                     12316 non-null  int64
11  Types_of_Junction                  12316 non-null  int64
12  Road_surface_type                  12316 non-null  int64
13  Road_surface_conditions            12316 non-null  int64
14  Light_conditions                   12316 non-null  int64
15  Weather_conditions                 12316 non-null  int64
16  Type_of_collision                  12316 non-null  int64
17  Vehicle_movement                   12316 non-null  int64
18  Casualty_class                     12316 non-null  int64
19  Sex_of_casualty                    12316 non-null  int64
20  Age_band_of_casualty               12316 non-null  int64
21  Casualty_severity                  12316 non-null  int64
22  Pedestrian_movement                12316 non-null  int64
23  Cause_of_accident                  12316 non-null  int64
dtypes: int64(24)
memory usage: 2.3 MB
```

Correlation

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

```
[36]: <Axes: >
```



```
[37]: df1.head()
```

```
[37]:   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]

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

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

```
[39]: (array([ 0.15822071,  8.91539214,  0.1431894 ,  0.17458477,  5.34534549,
               4.49967858,  1.07767124,  1.10426215,  3.61654037,  3.28161464,
               0.1319306 ,  3.08648691,  6.99480557,  0.61510308, 16.08282359,
               1.14934538, 10.09632283,  2.20071197,  3.2168602 ,  0.12594479,
               13.77841337,  0.20273788,  0.39747982,  3.19366551]),
       array([9.23937958e-01, 1.15890328e-02, 9.30908116e-01, 9.16409114e-01,
               6.90673790e-02, 1.05416165e-01, 5.83427189e-01, 5.75721597e-01,
               1.63937473e-01, 1.93823502e-01, 9.36163348e-01, 2.13686893e-01,
               3.02759144e-02, 7.35244973e-01, 3.21854237e-04, 5.62889079e-01,
               6.42112839e-03, 3.32752607e-01, 2.00201664e-01, 9.38969394e-01,
               1.01872169e-03, 9.03599597e-01, 8.19763078e-01, 2.02536988e-01]))
```

```
[40]: #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
```

```
[40]:
```

	features	Fscore	Pvalues
0	Day_of_week	0.158221	0.923938
1	Age_band_of_driver	8.915392	0.011589
2	Sex_of_driver	0.143189	0.930908
3	Educational_level	0.174585	0.916409
4	Vehicle_driver_relation	5.345345	0.069067
5	Driving_experience	4.499679	0.105416
6	Type_of_vehicle	1.077671	0.583427
7	Owner_of_vehicle	1.104262	0.575722
8	Area_accident_occured	3.616540	0.163937
9	Lanes_or_Medians	3.281615	0.193824
10	Road_allignment	0.131931	0.936163
11	Types_of_Junction	3.086487	0.213687
12	Road_surface_type	6.994806	0.030276
13	Road_surface_conditions	0.615103	0.735245
14	Light_conditions	16.082824	0.000322
15	Weather_conditions	1.149345	0.562889
16	Type_of_collision	10.096323	0.006421
17	Vehicle_movement	2.200712	0.332753
18	Casualty_class	3.216860	0.200202
19	Sex_of_casualty	0.125945	0.938969
20	Age_band_of_casualty	13.778413	0.001019
21	Casualty_severity	0.202738	0.903600
22	Pedestrian_movement	0.397480	0.819763
23	Cause_of_accident	3.193666	0.202537

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

```
[41]:
```

	features	Fscore	Pvalues
14	Light_conditions	16.082824	0.000322
20	Age_band_of_casualty	13.778413	0.001019
16	Type_of_collision	10.096323	0.006421
1	Age_band_of_driver	8.915392	0.011589
12	Road_surface_type	6.994806	0.030276
4	Vehicle_driver_relation	5.345345	0.069067
5	Driving_experience	4.499679	0.105416
8	Area_accident_occured	3.616540	0.163937
9	Lanes_or_Medians	3.281615	0.193824
18	Casualty_class	3.216860	0.200202
23	Cause_of_accident	3.193666	0.202537
11	Types_of_Junction	3.086487	0.213687

17	Vehicle_movement	2.200712	0.332753
15	Weather_conditions	1.149345	0.562889
7	Owner_of_vehicle	1.104262	0.575722
6	Type_of_vehicle	1.077671	0.583427
13	Road_surface_conditions	0.615103	0.735245
22	Pedestrian_movement	0.397480	0.819763
21	Casualty_severity	0.202738	0.903600
3	Educational_level	0.174585	0.916409
0	Day_of_week	0.158221	0.923938
2	Sex_of_driver	0.143189	0.930908
10	Road_allignment	0.131931	0.936163
19	Sex_of_casualty	0.125945	0.938969

we need higher Fscore and lower the Pvalues, so by evaluating, we can remove 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

```
[43]: #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()
```

```
[43]: 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	2	Going straight
1	2	2	Going straight
2	2	2	Going straight
3	2	2	Going straight
4	2	2	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

```
[44]: df2.shape
```

```
[44]: (12316, 17)
```

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12316 entries, 0 to 12315
Data columns (total 17 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Age_band_of_driver                    12316 non-null  object
1   Vehicle_driver_relation                12316 non-null  object
2   Driving_experience                     12316 non-null  object
3   Area_accident_occured                 12316 non-null  object
4   Lanes_or_Medians                      12316 non-null  object
5   Types_of_Junction                    12316 non-null  object
6   Road_surface_type                     12316 non-null  object
```

```

7   Light_conditions          12316 non-null object
8   Weather_conditions        12316 non-null object
9   Type_of_collision          12316 non-null object
10  Number_of_vehicles_involved 12316 non-null int64
11  Number_of_casualties        12316 non-null int64
12  Vehicle_movement           12316 non-null object
13  Casualty_class             12316 non-null object
14  Age_band_of_casualty        12316 non-null object
15  Cause_of_accident           12316 non-null object
16  Accident_severity           12316 non-null object
dtypes: int64(2), object(15)
memory usage: 1.6+ MB

```

```

[46]: #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']

```

```

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

```

```

Age_band_of_driver
18-30      4271
31-50      4087
Over 51     1585
Unknown    1548
Under 18    825
Name: count, dtype: int64
Vehicle_driver_relation
Employee    10206
Owner       1973
Other        123
Unknown       14
Name: count, dtype: int64
Driving_experience
5-10yr      4192
2-5yr       2613
Above 10yr   2262
1-2yr       1756
Below 1yr    1342
No Licence   118
unknown       33

```

Name: count, dtype: int64

Area_accident_occured

Other	4058
Office areas	3451
Residential areas	2060
Church areas	1060
Industrial areas	456
School areas	415
Recreational areas	327
Outside rural areas	218
Hospital areas	121
Market areas	63
Rural village areas	44
Unknown	22
Rural village areasOffice areas	20
Recreational areas	1

Name: count, dtype: int64

Lanes_or_Medians

Two-way (divided with broken lines road marking)	4796
Undivided Two way	3796
other	1660
Double carriageway (median)	1020
One way	845
Two-way (divided with solid lines road marking)	142
Unknown	57

Name: count, dtype: int64

Types_of_Junction

Y Shape	5430
No junction	3837
Crossing	2177
Other	445
Unknown	191
O Shape	164
T Shape	60
X Shape	12

Name: count, dtype: int64

Road_surface_type

Asphalt roads	11468
Earth roads	358
Gravel roads	242
Other	167
Asphalt roads with some distress	81

Name: count, dtype: int64

Light_conditions

Daylight	8798
Darkness - lights lit	3286
Darkness - no lighting	192
Darkness - lights unlit	40


```

Name: count, dtype: int64
Weather_conditions
Normal          10063
Raining         1331
Other           296
Unknown         292
Cloudy          125
Windy           98
Snow            61
Raining and Windy  40
Fog or mist     10
Name: count, dtype: int64
Type_of_collision
Vehicle with vehicle collision      8929
Collision with roadside objects    1786
Collision with pedestrians         896
Rollover                          397
Collision with animals             171
Collision with roadside-parked vehicles  54
Fall from vehicles                 34
Other                             26
Unknown                           14
With Train                        9
Name: count, dtype: int64
Vehicle_movement
Going straight      8466
Moving Backward     985
Other               937
Reversing           563
Turnover            489
Getting off         339
Entering a junction  193
Overtaking          96
Unknown             88
Stopping            61
U-Turn              50
Waiting to go       39
Parked              10
Name: count, dtype: int64
Casualty_class
Driver or rider     4944
na                  4443
Pedestrian          1649
Passenger           1280
Name: count, dtype: int64
Age_band_of_casualty
na                  4443
18-30              3145

```

```

31-50      2455
Under 18    1035
Over 51     994
5           244
Name: count, dtype: int64
Cause_of_accident
No distancing      2263
Changing lane to the right  1808
Changing lane to the left  1473
Driving carelessly  1402
No priority to vehicle  1207
Moving Backward    1137
No priority to pedestrian  721
Other              456
Overtaking         430
Driving under the influence of drugs  340
Driving to the left  284
Getting off the vehicle improperly  197
Driving at high speed  174
Overturning        149
Turnover           78
Overspeed          61
Overloading        59
Drunk driving       27
Unknown            25
Improper parking    25
Name: count, dtype: int64
Accident_severity
Slight Injury      10415
Serious Injury     1743
Fatal injury       158
Name: count, dtype: int64

```

```

[48]: 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()

```

```

[48]:   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_Unknown \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

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

	Vehicle_driver_relation_Unknown	Driving_experience_2-5yr \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	True

	Driving_experience_5-10yr	Driving_experience_Above 10yr ... \
0	False	False ...
1	False	True ...
2	False	False ...
3	True	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 102 columns]

```
[49]: df3=pd.concat([df2,dummy],axis=1)
df3.head()
```

```
[49]: 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]

```
[50]: #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()
```

```
[50]: Number_of_vehicles_involved Number_of_casualties Accident_severity \
0 2 2 Slight Injury
1 2 2 Slight Injury
2 2 2 Serious Injury
3 2 2 Slight Injury
4 2 2 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 18	Age_band_of_driver_Unknown	\
0	False	False	
1	False	False	
2	False	False	
3	False	False	
4	False	False	

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

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

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

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

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

[5 rows x 105 columns]

Seperating Independent and Dependent

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

```
[52]: (12316, 104)
```

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

```
[53]:   Number_of_vehicles_involved  Number_of_casualties  \
0                               2                      2
1                               2                      2
2                               2                      2
3                               2                      2
4                               2                      2

   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_Unknown  \
0                        False                        False
1                        False                        False
2                        False                        False
3                        False                        False
4                        False                        False

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

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

   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]

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

```
[54]: 0    Slight Injury
      1    Slight Injury
      2    Serious Injury
      3    Slight Injury
      4    Slight Injury
      Name: Accident_severity, dtype: object
```

```
[55]: y.value_counts()
```

```
[55]: Accident_severity
      Slight Injury      10415
      Serious Injury    1743
      Fatal injury      158
      Name: count, dtype: int64
```

Splitting the data

KNN Model Creation

Prediction

```
[64]: 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=42)
      from sklearn.neighbors import KNeighborsClassifier

      model_KNN = KNeighborsClassifier(n_neighbors=5)
      model_KNN.fit(x_train, y_train)
```

```
[64]: KNeighborsClassifier()
```

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

```
[67]: y_pred
```

```
[67]: array(['Slight Injury', 'Slight Injury', 'Slight Injury', ...,
      'Slight Injury', 'Slight Injury', 'Slight Injury'], dtype=object)
```

Checking Accuracy, Classification Report, Confusion Matrix

```
[ ]: from sklearn.metrics import
      ↪confusion_matrix,classification_report,,accuracy_score,ConfusionMatrixDisplay
```

Classification Report

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

Accuracy Score

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

Confusion Matrix

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