

## ASSIGNMENT-2 NAME: G.Prem Kumar REG NO:21BCE9733

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
dataset=pd.read_csv("car_crashes.csv")
dataset
```

	total	speeding	alcohol	not_distracted	no_previous	ins_premium
	18.8	7.332	5.640	18.048	15.040	784.55
\						
0						
1	18.1	7.421	4.525	16.290	17.014	
1053.48						
2	18.6	6.510	5.208	15.624	17.856	
899.47	3	22.4	4.032	5.824	21.056	21.280
827.34						
4	12.0	4.200	3.360	10.920	10.680	
878.41						
5	13.6	5.032	3.808	10.744	12.920	
835.50	6	10.8	4.968	3.888	9.396	8.856
1068.73	7	16.2	6.156	4.860	14.094	16.038
1137.87	8	5.9	2.006	1.593	5.900	5.900
1273.89	9	17.9	3.759	5.191	16.468	16.826
1160.13						
10	15.6	2.964	3.900	14.820	14.508	913.15
11	17.5	9.450	7.175	14.350	15.225	861.18
12	15.3	5.508	4.437	13.005	14.994	641.96
13	12.8	4.608	4.352	12.032	12.288	803.11
14	14.5	3.625	4.205	13.775	13.775	710.46
15	15.7	2.669	3.925	15.229	13.659	649.06

16	17.8	4.806	4.272	13.706	15.130	780.45
17	21.4	4.066	4.922	16.692	16.264	872.51
18	20.5	7.175	6.765	14.965	20.090	1281.55

19	15.1	5.738	4.530	13.137	12.684	661.88
20	12.5	4.250	4.000	8.875	12.375	1048.78
21	8.2	1.886	2.870	7.134	6.560	1011.14
22	14.1	3.384	3.948	13.395	10.857	1110.61
23	9.6	2.208	2.784	8.448	8.448	777.18
24	17.6	2.640	5.456	1.760	17.600	896.07
25	16.1	6.923	5.474	14.812	13.524	790.32
26	21.4	8.346	9.416	17.976	18.190	816.21
27	14.9	1.937	5.215	13.857	13.410	732.28
28	14.7	5.439	4.704	13.965	14.553	1029.87
29	11.6	4.060	3.480	10.092	9.628	746.54
30	11.2	1.792	3.136	9.632	8.736	1301.52
31	18.4	3.496	4.968	12.328	18.032	869.85
32	12.3	3.936	3.567	10.824	9.840	1234.31
33	16.8	6.552	5.208	15.792	13.608	708.24
34	23.9	5.497	10.038	23.661	20.554	688.75
35	14.1	3.948	4.794	13.959	11.562	697.73
36	19.9	6.368	5.771	18.308	18.706	881.51
37	12.8	4.224	3.328	8.576	11.520	804.71
38	18.2	9.100	5.642	17.472	16.016	905.99
39	11.1	3.774	4.218	10.212	8.769	1148.99
40	23.9	9.082	9.799	22.944	19.359	858.97
41	19.4	6.014	6.402	19.012	16.684	669.31
42	19.5	4.095	5.655	15.990	15.795	767.91
43	19.4	7.760	7.372	17.654	16.878	1004.75
44	11.3	4.859	1.808	9.944	10.848	809.38



45	13.6	4.080	4.080	13.056	12.920	716.20
46	12.7	2.413	3.429	11.049	11.176	768.95
47	10.6	4.452	3.498	8.692	9.116	890.03
48	23.8	8.092	6.664	23.086	20.706	992.61
49	13.8	4.968	4.554	5.382	11.592	670.31
50	17.4	7.308	5.568	14.094	15.660	791.14
	ins_losses abbrev					
0	145.08					
	AL					
1	133.93					
	AK					
2	110.35					
	AZ					
3	142.39					
	AR					
4	165.63					
	CA					
5	139.91					
	CO					
6	167.02					
	CT					
7	151.48					
	DE					
8	136.05					
	DC					
9	144.18					
	FL					
10	142.80					
	GA					
11	120.92					
	HI					
12	82.75	ID				
13	139.15					
	IL					
14	108.92					
	IN					
15	114.47					
	IA					
16	133.80					
	KS					
17	137.13					
	KY					
18	194.78					
	LA					
19	96.57	ME				

20	192.70	
	MD	
21	135.63	
	MA	
22	152.26	
	MI	
23	133.35	
	MN	
24	155.77	
	MS	
25	144.45	
	MO	
26	85.15	MT
27	114.82	
	NE	
28	138.71	
	NV	
29	120.21	
	NH	
30	159.85	
	NJ	
31	120.75	
	NM	
32	150.01	
	NY	
33	127.82	
	NC	
34	109.72	
	ND	



35	133.52	
	OH	
36	178.86	
	OK	
37	104.61	
	OR	
38	153.86	
	PA	
39	148.58	
	RI	
40	116.29	
	SC	
41	96.87	SD
42	155.57	
	TN	
43	156.83	
	TX	
44	109.48	
	UT	
45	109.61	
	VT	
46	153.72	
	VA	
47	111.62	
	WA	
48	152.56	
	WV	
49	106.62	
	WI	
50	122.04	
	WY	

dataset.info()

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

RangeIndex: 51 entries, 0 to 50 Data

columns (total 8 columns):

#	Column	Non-Null Count	Dtype	---
0	total	51 non-null	float64	
1	speeding	51 non-null	float64	2 alcohol
51	non-null	float64		
3	not_distracted	51 non-null	float64	
4	no_previous	51 non-null	float64	
5	ins_premium	51 non-null	float64	



```
6      ins_losses      51 non-null      float64 7      abbrev      51
non-null      object dtypes: float64(7), object(1) memory usage: 3.3+
KB dataset.head(8)
```

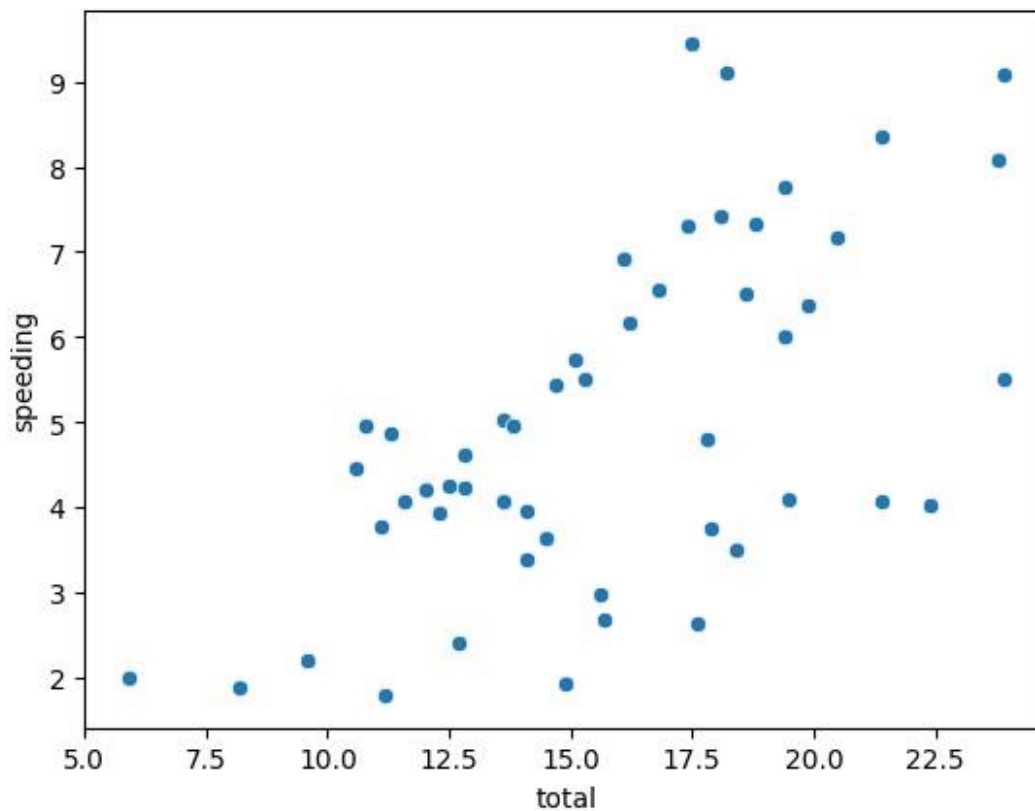
```
      total  speeding  alcohol  not_distracted  no_previous  ins_premium
0
1  18.1      7.421    4.525      16.290      17.014    1053.48 2
18.6      6.510    5.208      15.624      17.856      899.47
3  22.4      4.032    5.824      21.056      21.280      827.34
4  12.0      4.200    3.360      10.920      10.680      878.41
5  13.6      5.032    3.808      10.744      12.920      835.50
```

6	10.8	4.968	3.888	9.396	8.856	1068.73
7	16.2	6.156	4.860	14.094	16.038	1137.87

	ins_losses	abbrev
0	145.08	AL
1	133.93	AK
2	110.35	AZ
3	142.39	AR
4	165.63	CA
5	139.91	CO
6	167.02	CT
7	151.48	DE

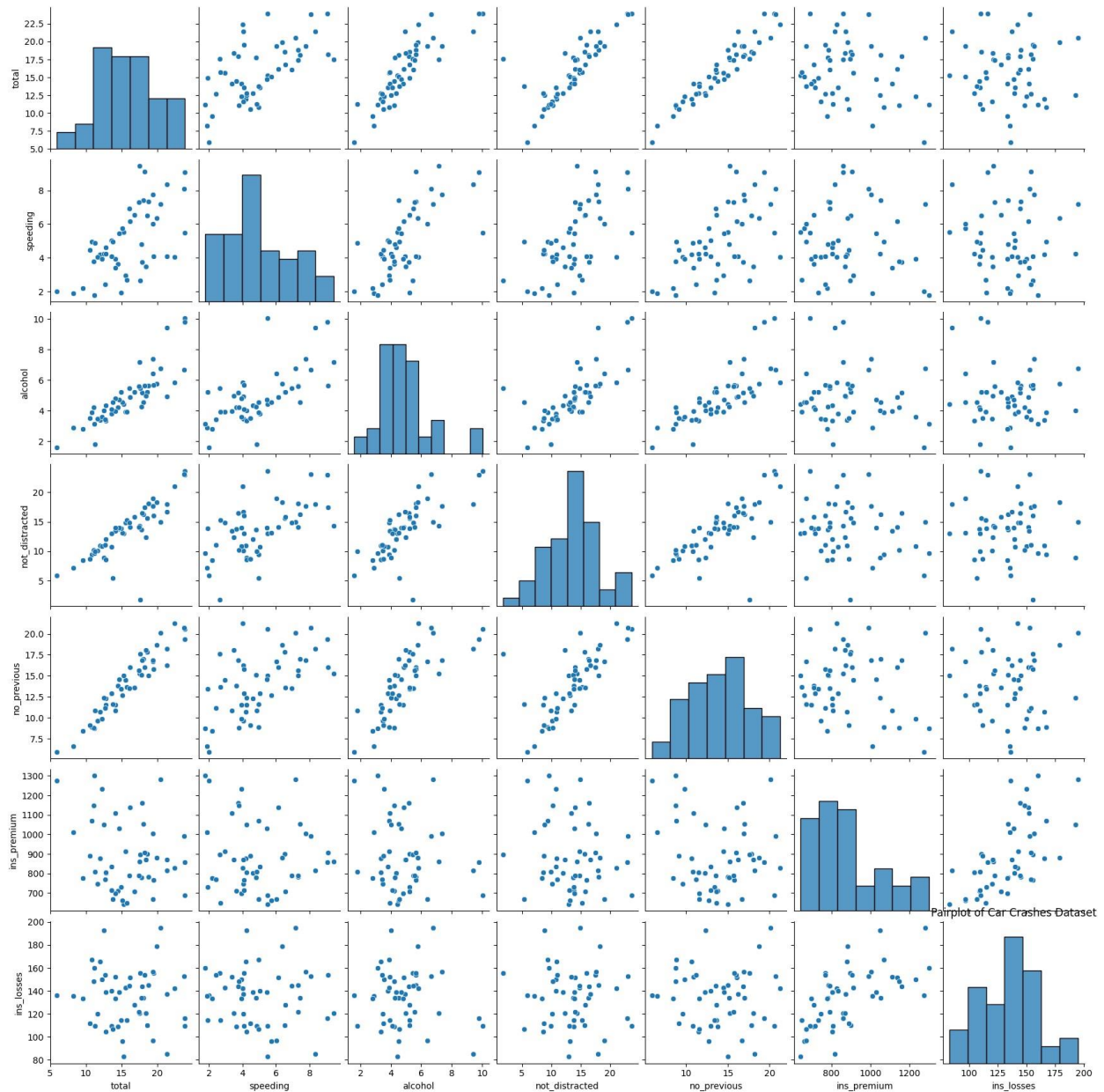
```
sns.scatterplot(x="total", y="speeding", data=dataset)
```

```
<Axes: xlabel='total', ylabel='speeding'>
```



```
# Inference: from the plot we can say that as the total increases
speeding is decreases
```

```
sns.pairplot(dataset)
plt.title("Pairplot of Car Crashes Dataset") plt.show()
```



*# Inference: The pairplot provides a quick overview of the relationships between numeric variables in the dataset. It helps identify potential correlations or patterns.*

```
sns.distplot(dataset["total"], bins=20, kde=True)
plt.title("Histogram of Total Number of Accidents")
plt.xlabel("Total Accidents")
```

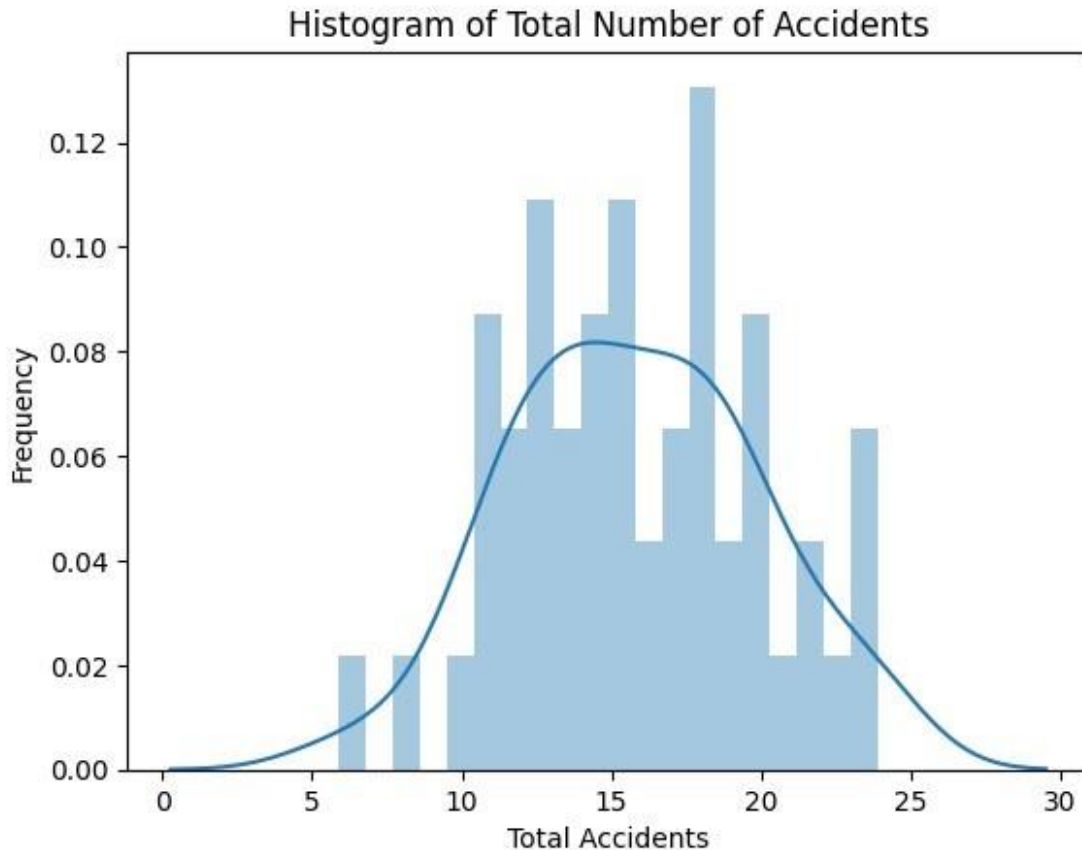
```
plt.ylabel("Frequency")
plt.show()
```

```
<ipython-input-24-c2887f4da83f>:1: UserWarning:
`distplot` is a deprecated function and will be removed in seaborn
v0.14.0.
```

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

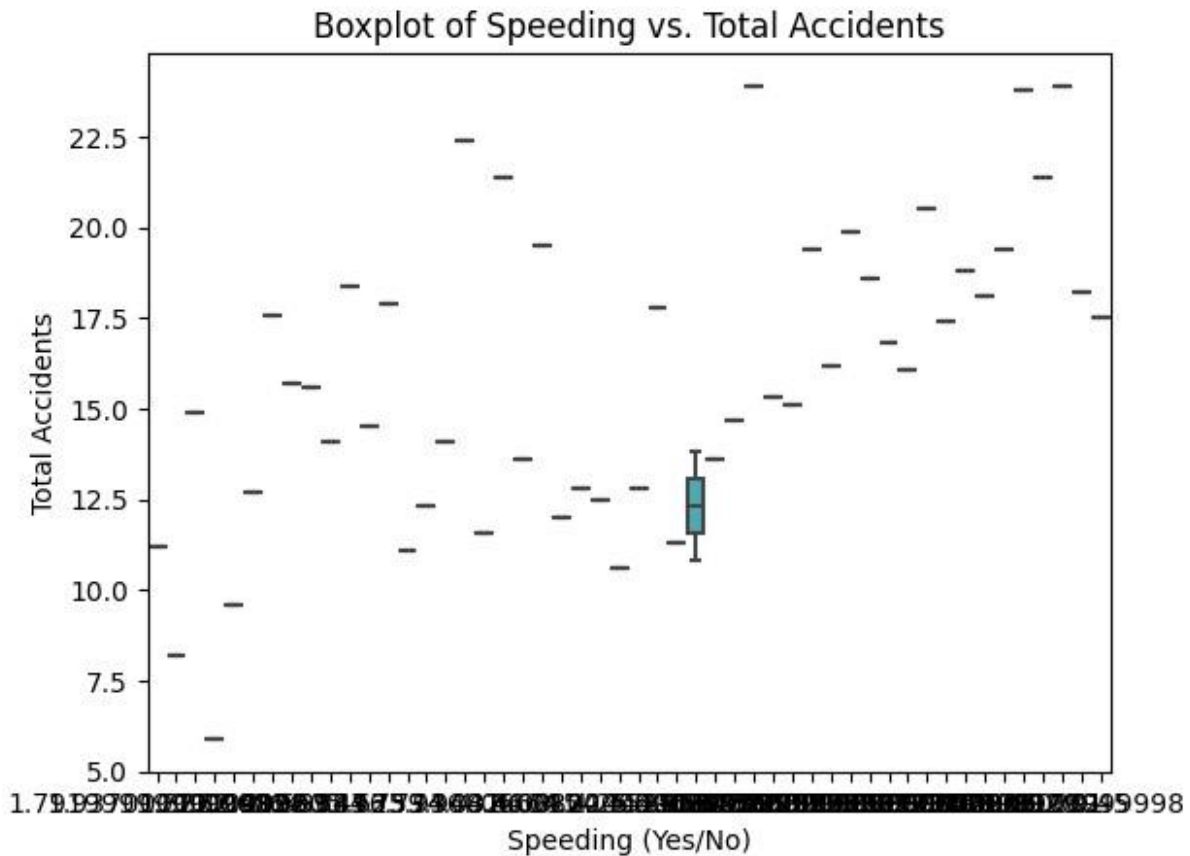
For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(dataset["total"], bins=20, kde=True)
```



*# Inference: The histogram shows the distribution of total accidents. Most states have a relatively low number of accidents, with a few outliers with significantly higher accident counts.*

```
sns.boxplot(x="speeding", y="total", data=dataset) plt.title("Boxplot of Speeding vs. Total Accidents") plt.xlabel("Speeding (Yes/No)") plt.ylabel("Total Accidents") plt.show()
```

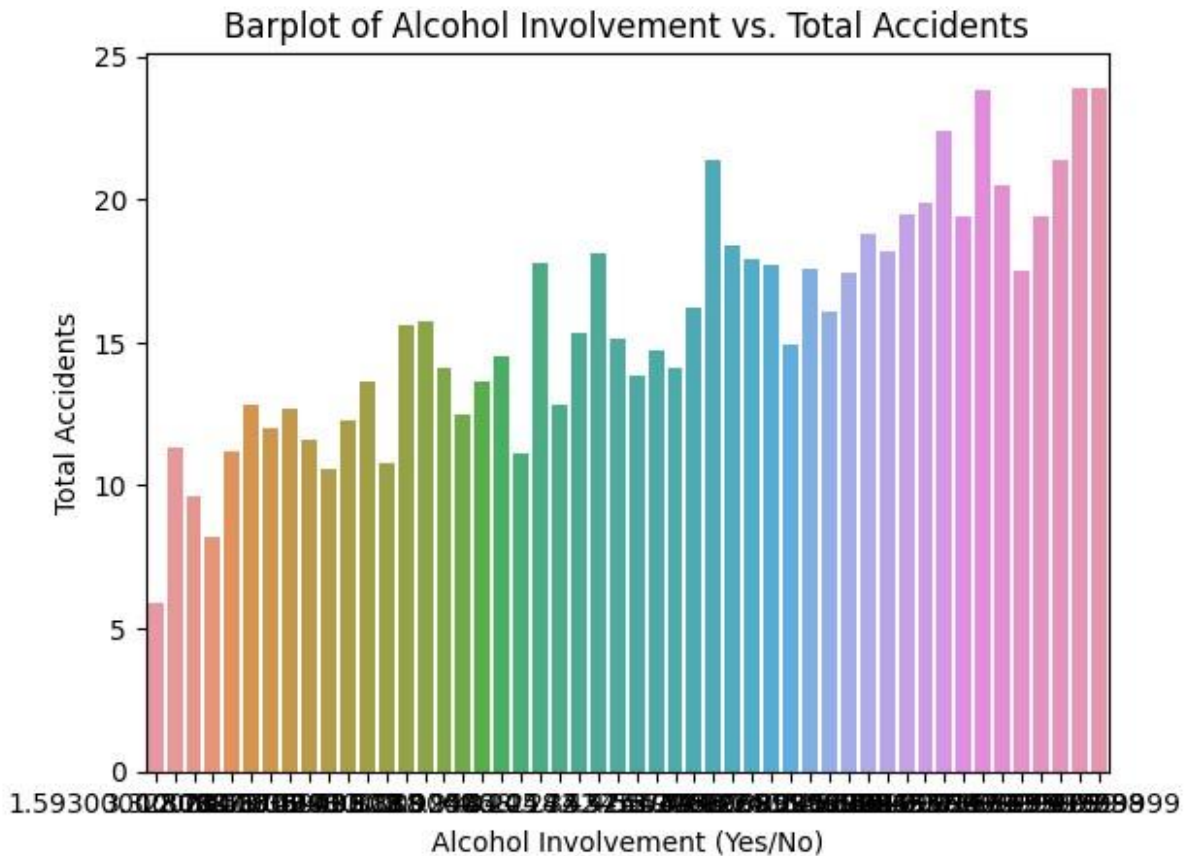


*# Inference: The boxplot illustrates the relationship between speeding (yes/no) and the total number of accidents. It indicates that states with higher speeding rates tend to have a higher median total number of accidents.*

```
sns.barplot(x="alcohol", y="total", data=dataset, ci=None)
plt.title("Barplot of Alcohol Involvement vs. Total Accidents")
plt.xlabel("Alcohol Involvement (Yes/No)") plt.ylabel("Total
Accidents") plt.show()
```

<ipython-input-19-e9d4c62a021d>:1: FutureWarning:  
The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

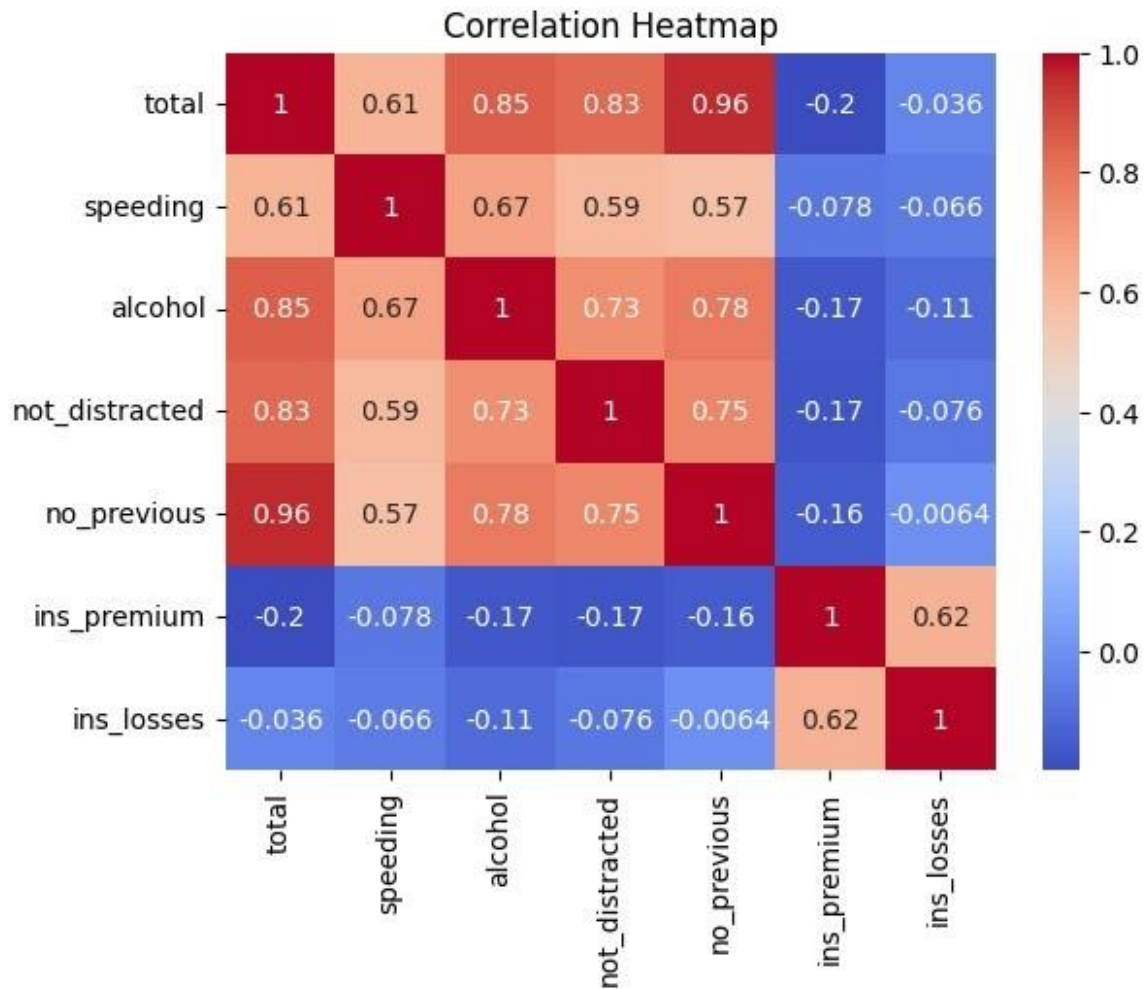
```
sns.barplot(x="alcohol", y="total", data=dataset, ci=None)
```



*# Inference: The barplot compares the total number of accidents for states with and without alcohol involvement. It suggests that states with alcohol involvement tend to have a higher average number of accidents.*

```
correlation_matrix = dataset.corr()
sns.heatmap(correlation_matrix, annot=True, cmap="coolwarm")
plt.title("Correlation Heatmap") plt.show()
```

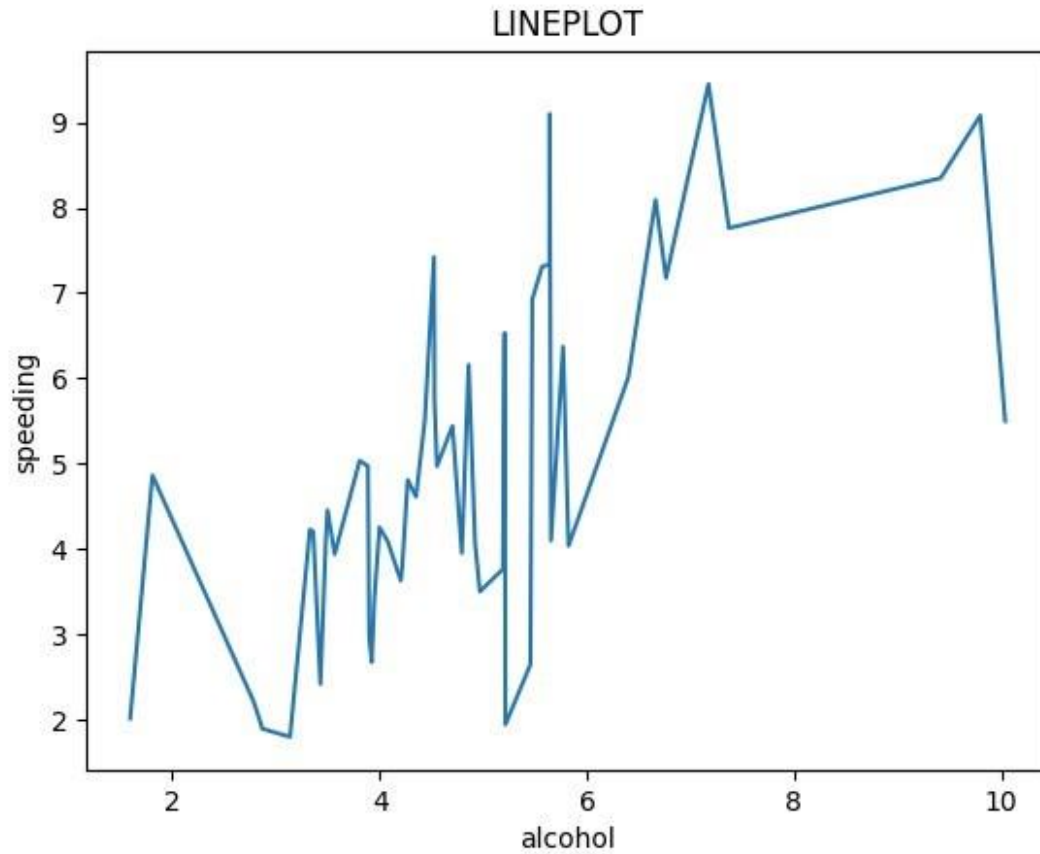
<ipython-input-21-f966e5b914d1>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning. correlation\_matrix = dataset.corr()



*# Inference: The heatmap displays the correlation between numeric variables in the dataset. Positive correlations are shown in warmer colors, while negative correlations are in cooler colors. It helps identify potential relationships between variables.*

```
sns.lineplot(x="alcohol",y="speeding",data=dataset)
plt.title("LINEPLOT")
```

```
Text(0.5, 1.0, 'LINEPLOT')
```

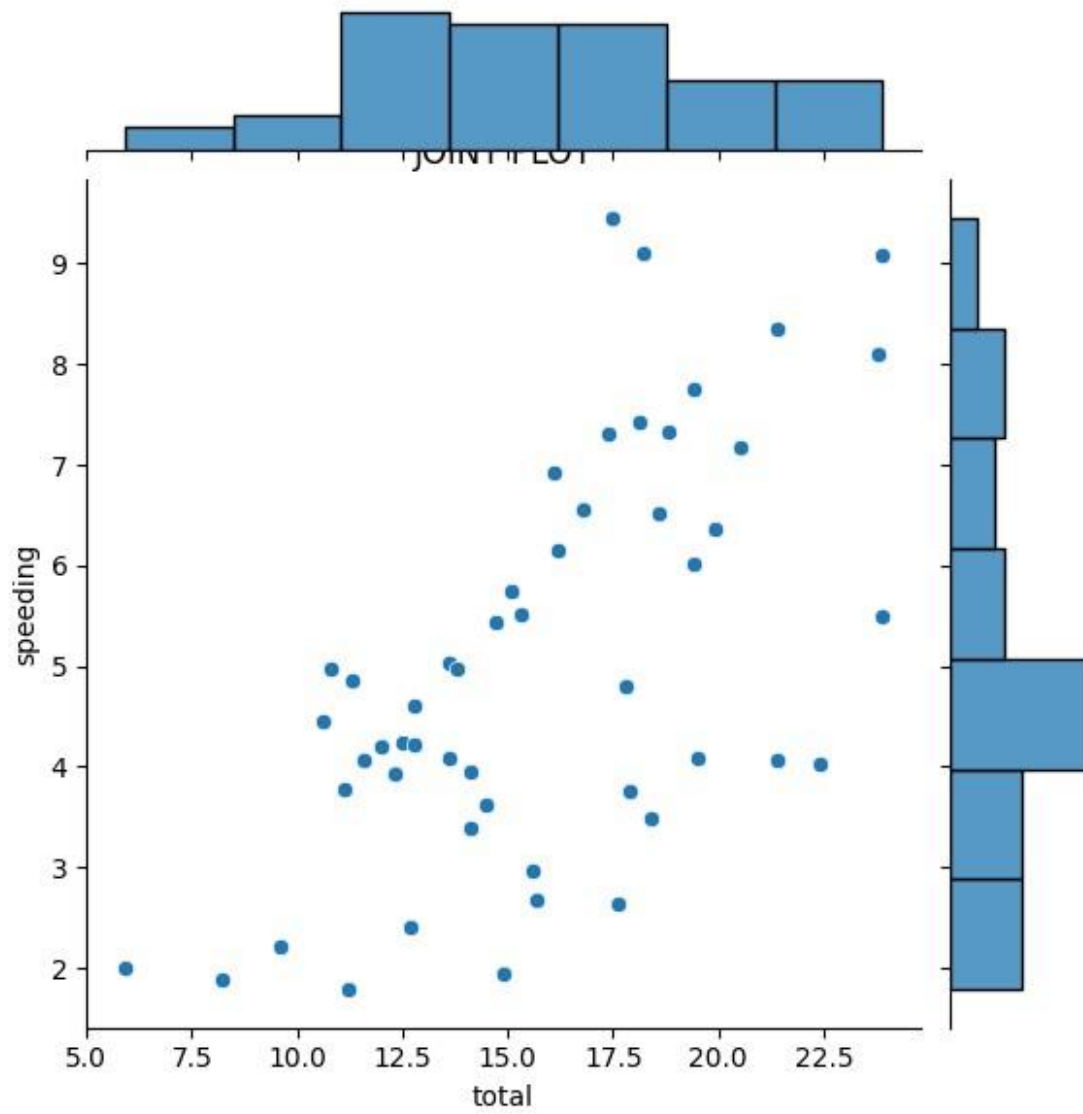


*# Inference: The line plot comparing "Alcohol" and "Speeding" incidents in car crashes shows that alcohol with higher value have higher speeding value.*

```
sns.jointplot(x="total",y="speeding",data=dataset) plt.title("JOINT")
```

```
Text(0.5, 1.0, 'JOINT PLOT')
```





*# INFERENCE :States with a higher rate of "Speeding" incidents tend to have a wider range of total accidents, as indicated by the larger interquartile range (IQR) and the presence of outliers.*