

```
print("Shubh Udaybhai Pandya, 21BCE5770")
```

```
Shubh Udaybhai Pandya, 21BCE5770
```

```
import seaborn as sns
```

```
print(sns.get_dataset_names())
```

```
['anagrams', 'anscombe', 'attention', 'brain_networks', 'car_crashes',  
'diamonds', 'dots', 'dowjones', 'exercise', 'flights', 'fmri', 'geyser',  
'glue', 'healthexp', 'iris', 'mpg', 'penguins', 'planets', 'seaice', 'taxis',  
'tips', 'titanic']
```

```
df=sns.load_dataset('car_crashes')
```

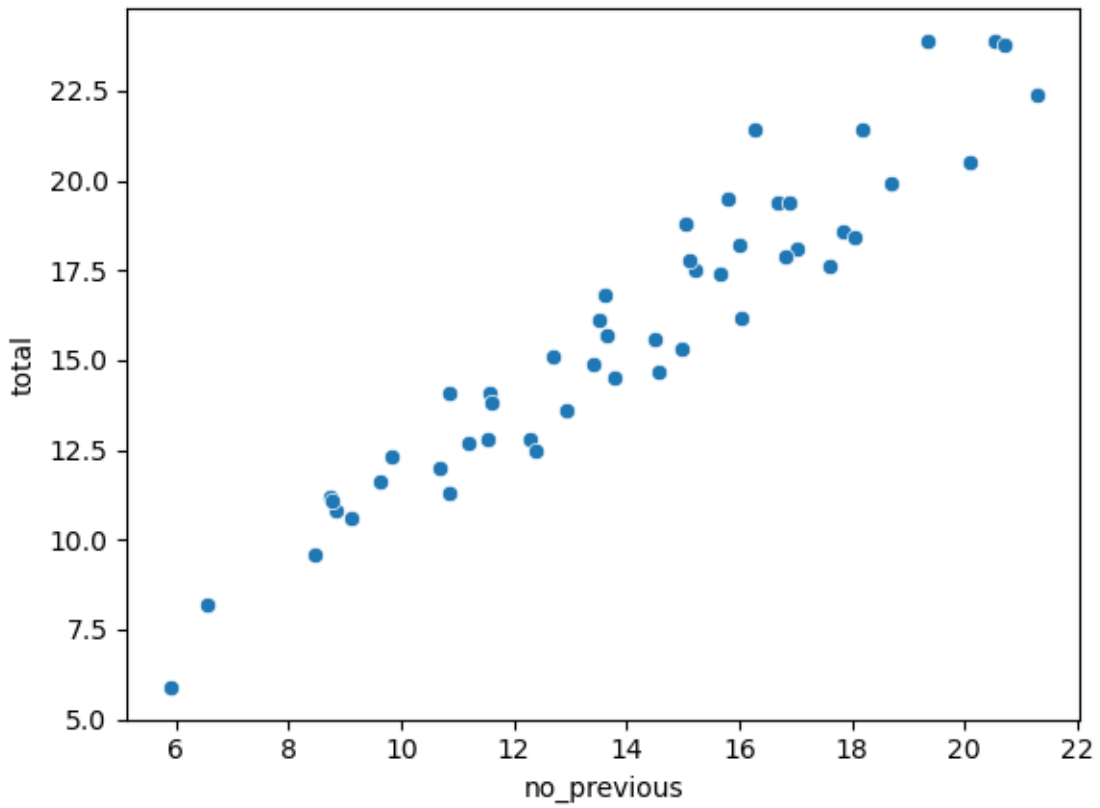
```
df.head(5)
```

	total	speeding	alcohol	not_distracted	no_previous	ins_premium \
0	18.8	7.332	5.640	18.048	15.040	784.55
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

	ins_losses	abbrev
0	145.08	AL
1	133.93	AK
2	110.35	AZ
3	142.39	AR
4	165.63	CA

```
sns.scatterplot(x="no_previous",y="total",data=df)
```

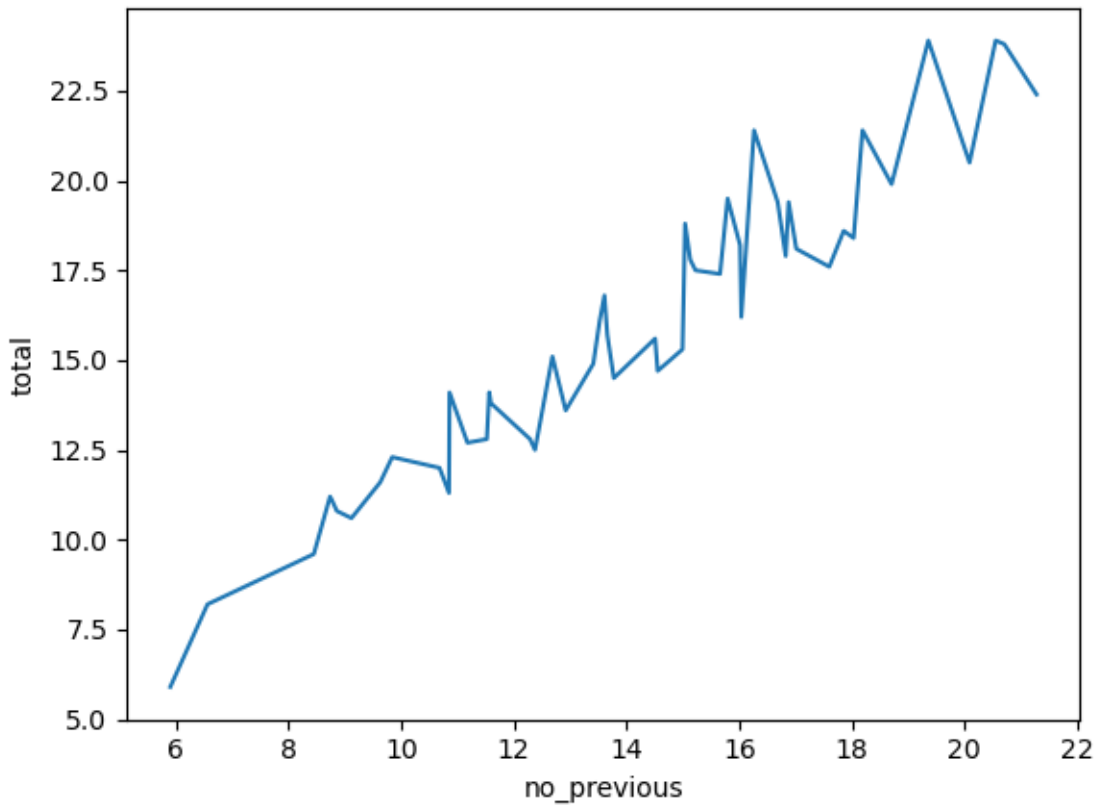
```
<Axes: xlabel='no_previous', ylabel='total'>
```



Inference: As the value of no_previous is increasing, the value of total is also increasing.

```
sns.lineplot(x="no_previous",y="total",data=df)
```

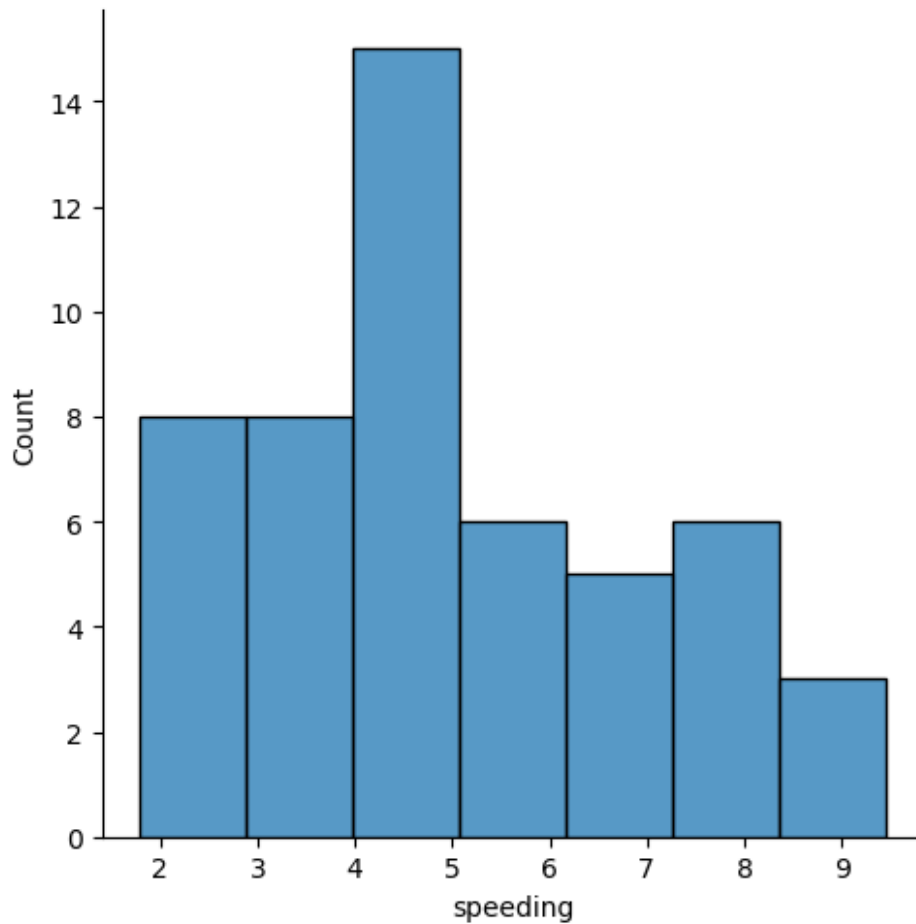
```
<Axes: xlabel='no_previous', ylabel='total'>
```



Inference: The line plot reveals an overall positive correlation between the "no_previous" and "total" variables, but there are instances where negative correlations exist, suggesting that in some cases, an increase in "no_previous" incidents might lead to a decrease in total outcomes.

```
sns.displot(df['speeding'])
```

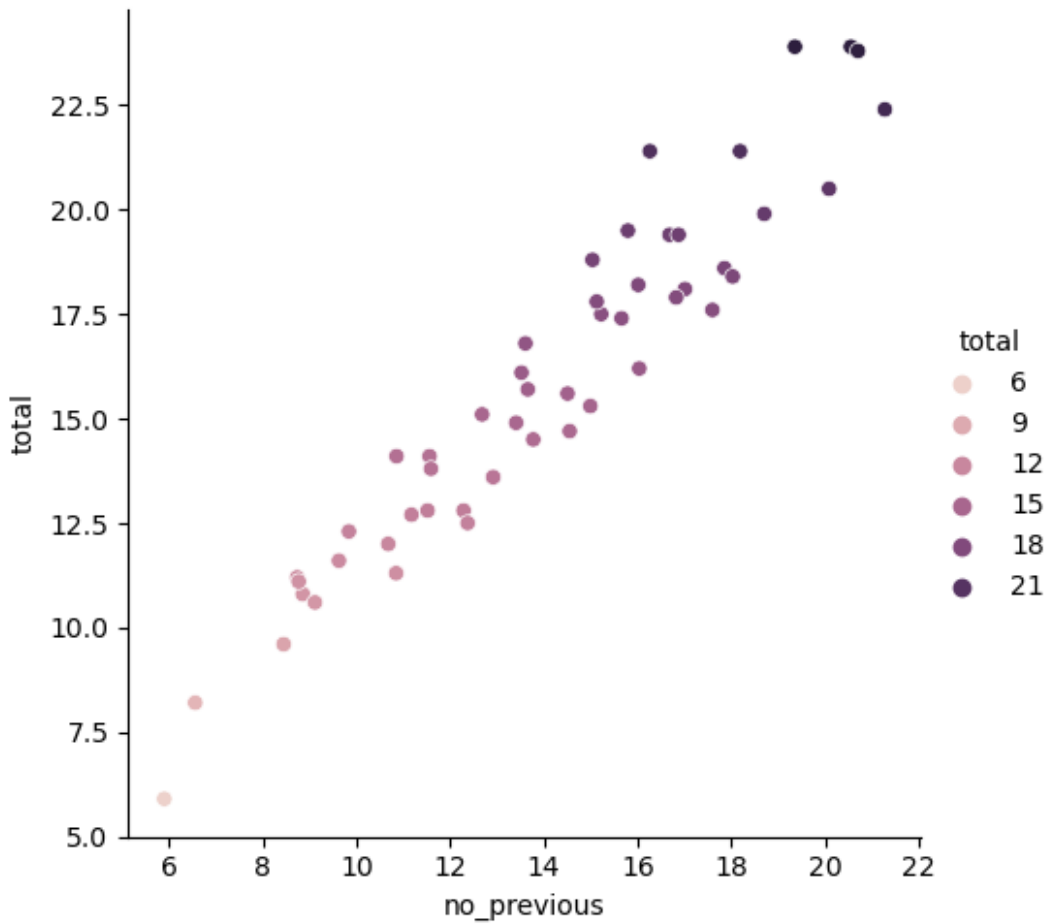
```
<seaborn.axisgrid.FacetGrid at 0x2b6c6480fd0>
```



Inference: The most number of speedings are between the value of 4-5. Most areas have low speeding rates, there are specific locations or cases with significantly higher rates of speeding.

```
sns.relplot(x="no_previous", y="total", data=df, hue="total")
```

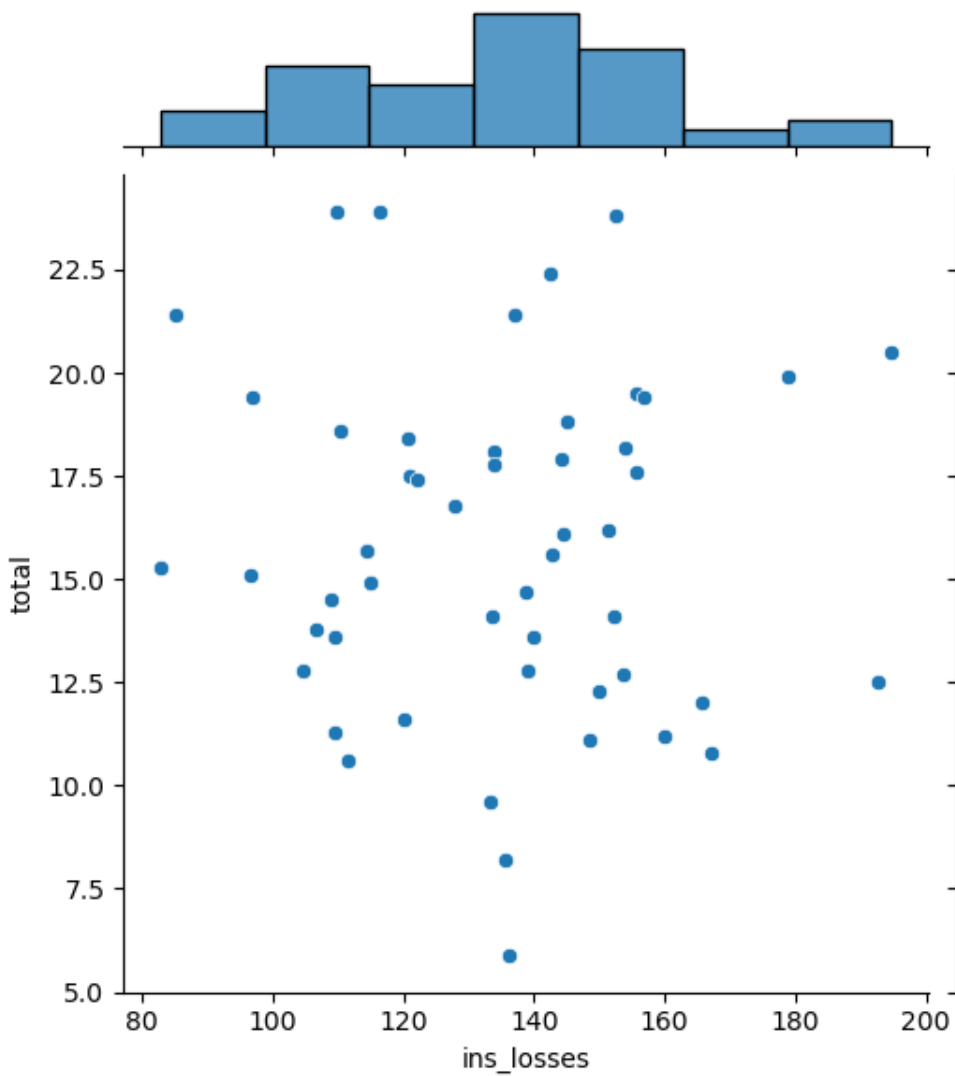
```
<seaborn.axisgrid.FacetGrid at 0x2b6d33c7790>
```



Inference: The relational plot shows a positive correlation between the number of incidents . However, there are variations in total outcomes for the same number of incidents, suggesting the influence of other factors on the results.

```
sns.barplot(data=df,x="speeding", y="alcohol")
```

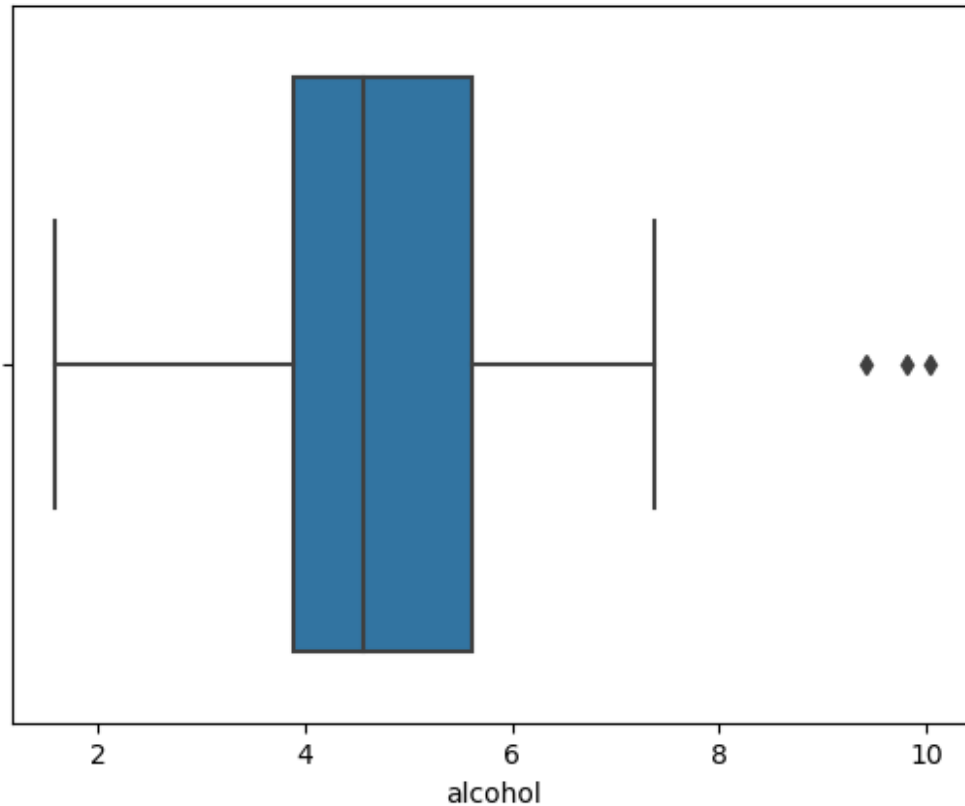
```
<Axes: xlabel='speeding', ylabel='alcohol'>
```

Inference: there are some regions where higher insurance losses seem to be associated with higher total crash numbers.

```
sns.boxplot(x="alcohol",data=df)
```

```
<Axes: xlabel='alcohol'>
```



Inference :The median appears to be around 5.0, suggesting that the median alcohol-related accidents are moderate. There are a few outliers on the right side of the plot, indicating that some observations have higher alcohol-related accident rates compared to the majority.

```
corr=df.corr()
```

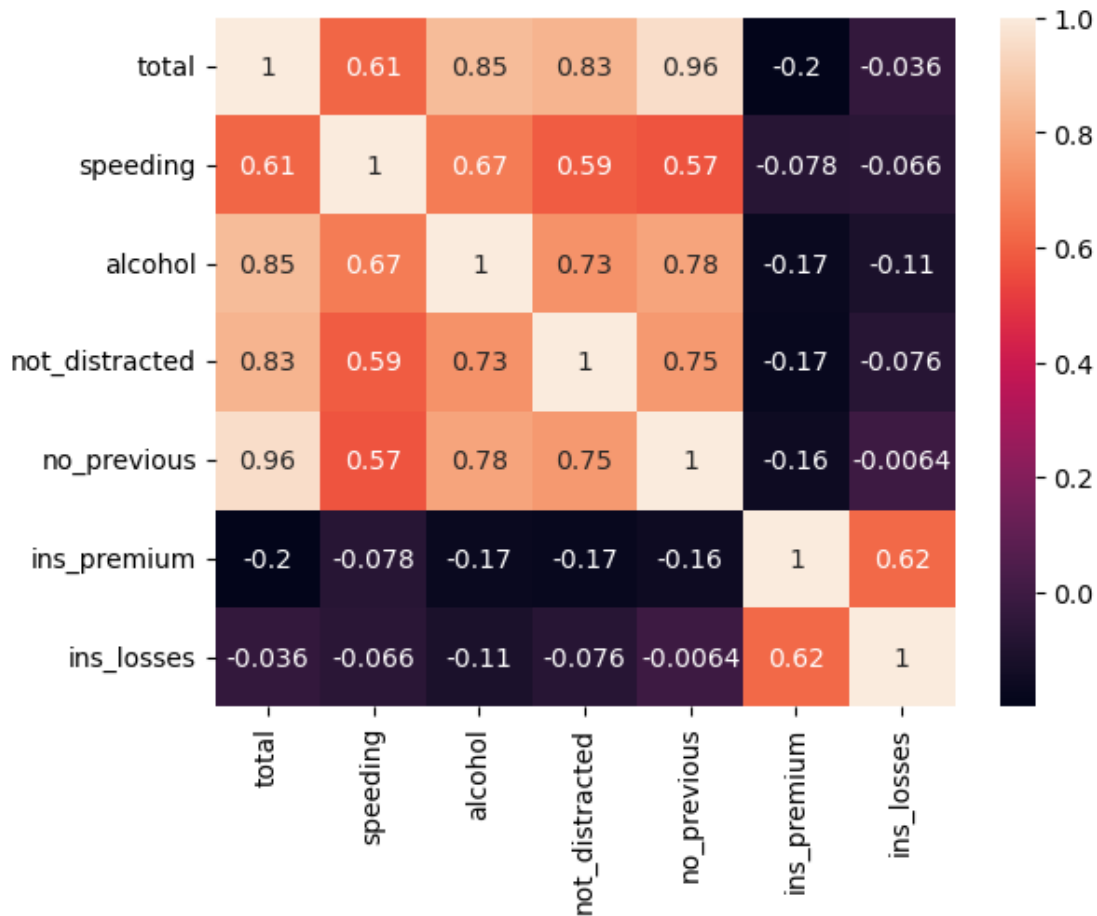
```
C:\Users\rajes\AppData\Local\Temp\ipykernel_1020\1726683880.py:1:
```

```
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.
```

```
corr=df.corr()
```

```
sns.heatmap(corr,annot=True)
```

```
<Axes: >
```

Inference: Each cell represents the correlation between two variables. The color intensity of each cell indicates the strength and direction of the correlation. Positive correlations are represented with light colors. Variables with a strong positive correlation will have higher values, often closer to 1 and vice versa. For eg- no_previous and total have highest positive correlation between them. Alcohol and non-distracted have highest negative correlation between them.