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

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

```
['anagrams', 'anscombe', 'attention', 'brain_networks', 'car_crashes', 'diamonds', 'dots', 'dowjones', 'exercise', 'flights', 'fmri', 'g
```

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

```
    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
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
```

```
sns.__version__
```

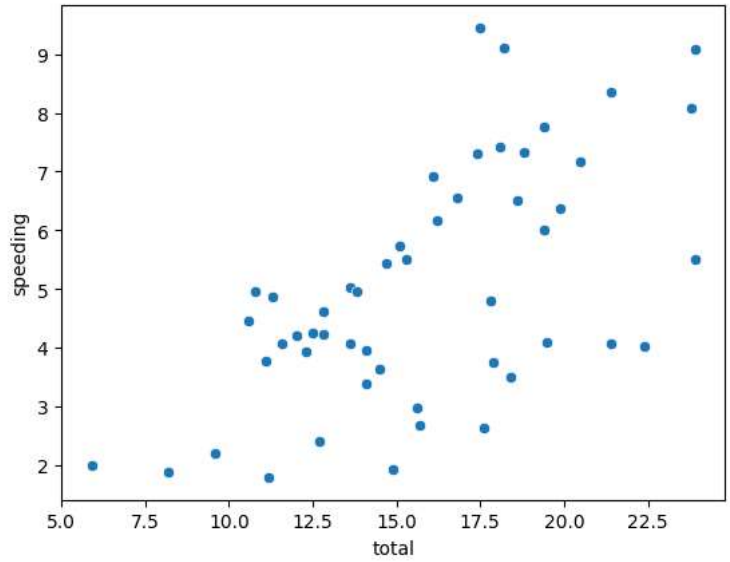
```
'0.12.2'
```

```
df.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
```

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

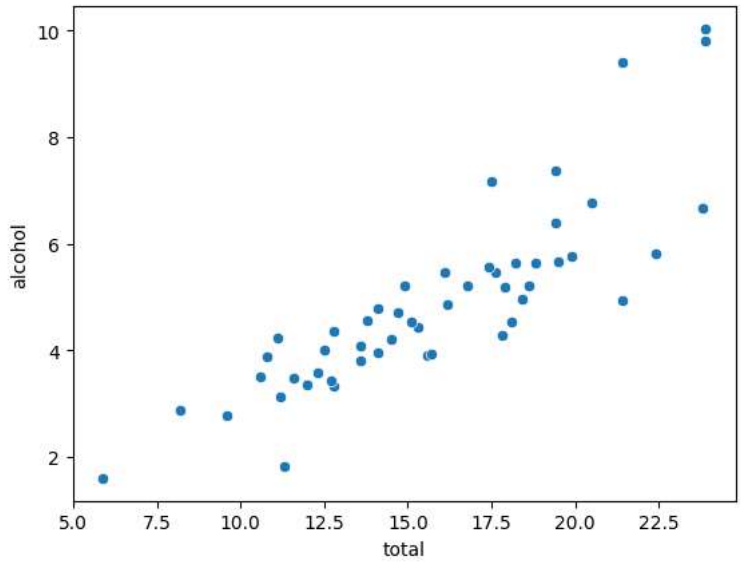
```
<Axes: xlabel='total', ylabel='speeding'>
```



Inference: Here we can clearly observe that increase in speed has caused many car crashes. Which shows the direct proportionality in between total and speeding of car crashes.

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

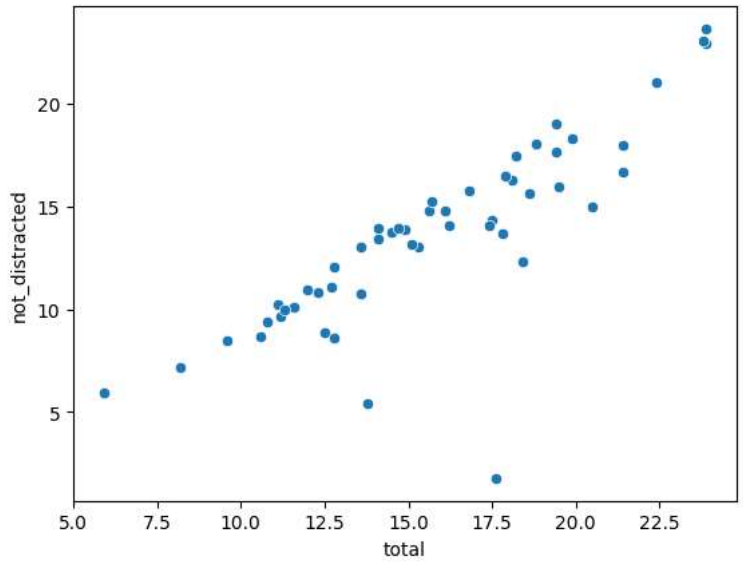
```
<Axes: xlabel='total', ylabel='alcohol'>
```



Inference: In the above graph ,it is obvious that excessive alcohol consumption has contributed to numerous auto accidents. which demonstrates the direct correlation between the overall number of crashes and alcohol use.

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

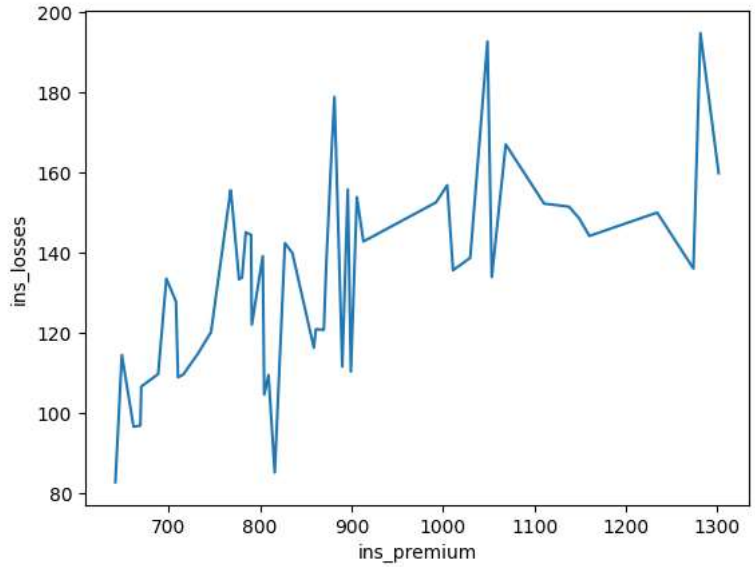
```
<Axes: xlabel='total', ylabel='not_distracted'>
```



Inference:Here we can clearly observe that not_distracted has caused many car crashes. Which shows the direct proportionality in between total and not_distracted of car crashes.

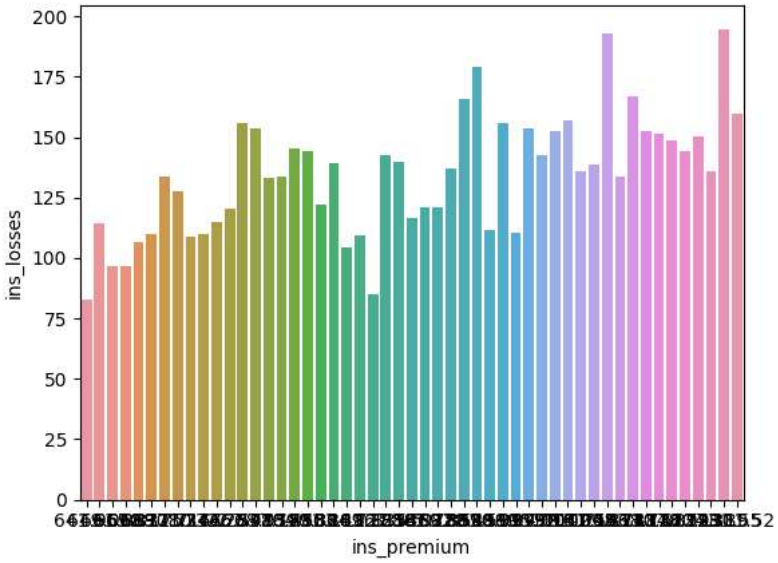
```
sns.lineplot(x="ins_premium",y="ins_losses",data=df)
```

```
<Axes: xlabel='ins_premium', ylabel='ins_losses'>
```



```
sns.barplot(data=df,x="ins_premium",y="ins_losses")
```

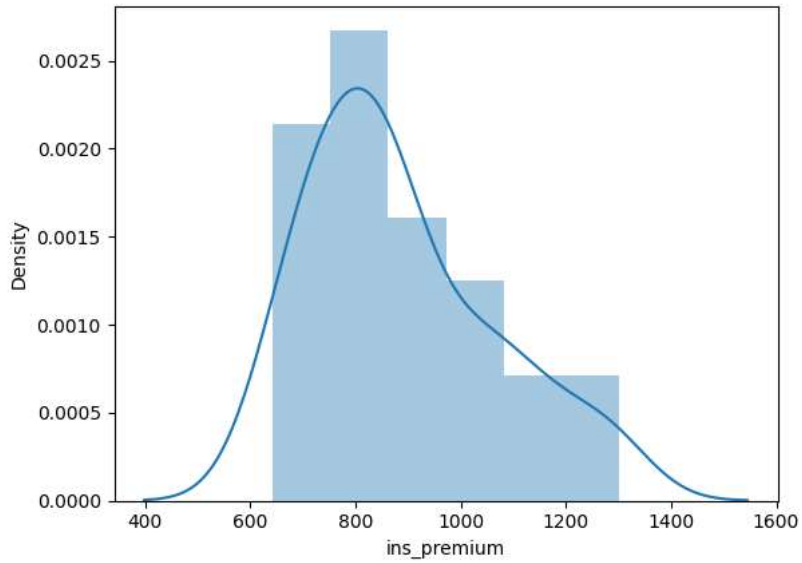
```
<Axes: xlabel='ins_premium', ylabel='ins_losses'>
```



Inference:There is an irregular plotting of lineplot of ins_premium and ins_losses. which defines that insurance losses is directly proportional to insurance premium. where increse in insurance premium also tends to increase in insurance loss.

```
sns.distplot(df['ins_premium'])
```

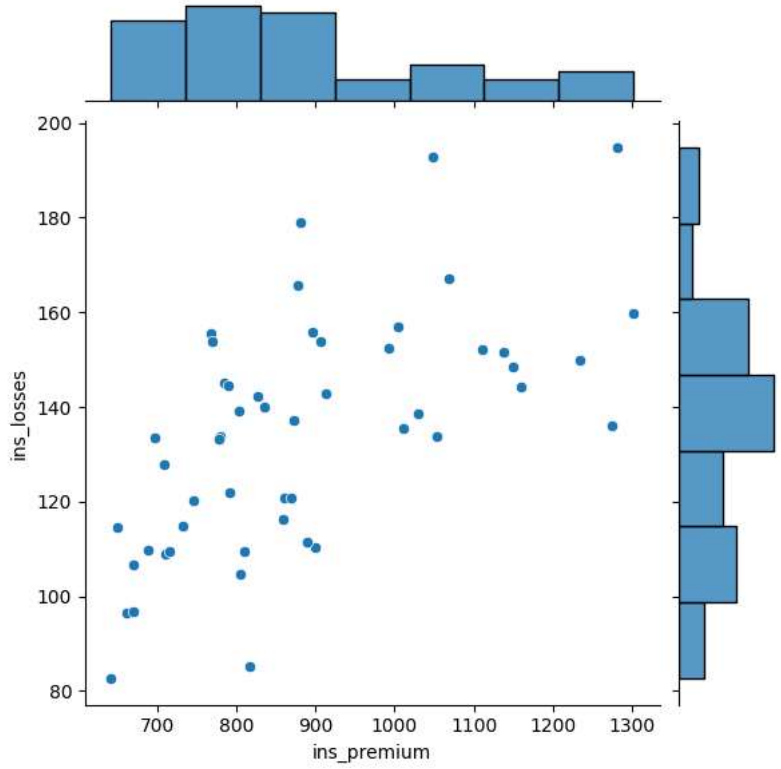
```
<ipython-input-4-2616b6ce983b>:1: UserWarning:

`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(df['ins_premium'])
<Axes: xlabel='ins_premium', ylabel='Density'>
```



Inference: The above graph is about the density of the insurance premium where there is a huge increase in the density of insurance premium at a point of 800.

```
sns.jointplot(data=df,x="ins_premium",y="ins_losses")
```

```
<seaborn.axisgrid.JointGrid at 0x7d1d5bfbbe50>
```



Inference: The above graph is about jointplot of ins_premium and ins_losses where the plotting of graph is randomly distributed through out the graph. We can observe the density of ins_premium and ins_losses where they show in great increase and decrease.

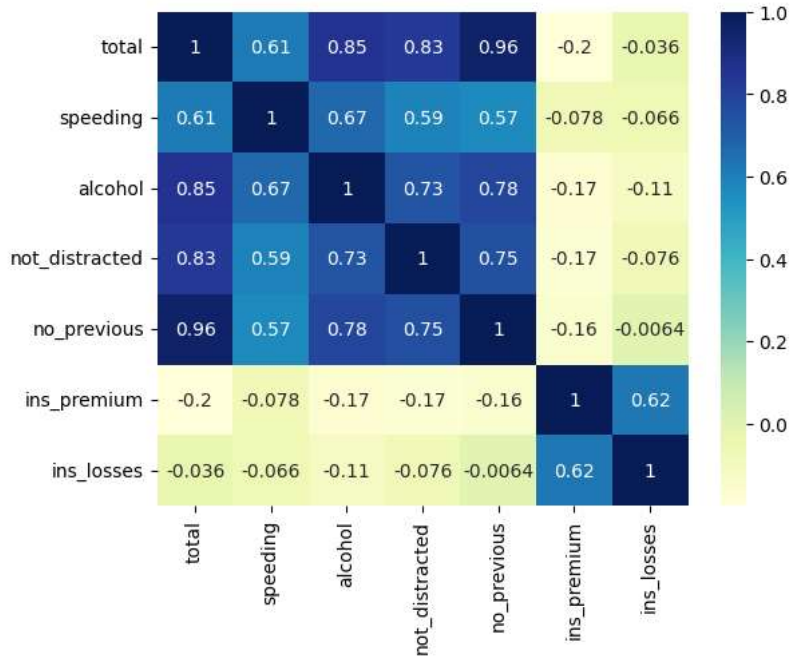
```
corr=df.corr()
corr
```

```
<ipython-input-22-7d5195e2bf4d>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future versior
  corr=df.corr()
```

	total	speeding	alcohol	not_distracted	no_previous	ins_premium	ins_losses
total	1.000000	0.611548	0.852613	0.827560	0.956179	-0.199702	-0.036011
speeding	0.611548	1.000000	0.669719	0.588010	0.571976	-0.077675	-0.065928
alcohol	0.852613	0.669719	1.000000	0.732816	0.783520	-0.170612	-0.112547
not_distracted	0.827560	0.588010	0.732816	1.000000	0.747307	-0.174856	-0.075970
no_previous	0.956179	0.571976	0.783520	0.747307	1.000000	-0.156895	-0.006359
ins_premium	-0.199702	-0.077675	-0.170612	-0.174856	-0.156895	1.000000	0.623116
ins_losses	-0.036011	-0.065928	-0.112547	-0.075970	-0.006359	0.623116	1.000000

```
sns.heatmap(corr,annot=True,cmap="YlGnBu")
```

```
<Axes: >
```



Inferance: The above data is the correlation of the car crashs data. Where we can observe the clear correaltion between each and every info of the car crashs data.