```
import seaborn as sns
df=sns.load dataset("car crashes")
df
                                not_distracted
                                                 no_previous
    total
           speeding
                      alcohol
                                                               ins premium
     18.8
                                         18.048
                                                                     784.55
               7.332
                        5.640
                                                       15.040
     18.1
               7.421
                        4.525
                                         16.290
                                                       17.014
                                                                    1053.48
1
2
     18.6
               6.510
                        5.208
                                         15.624
                                                       17.856
                                                                     899.47
     22.4
3
               4.032
                        5.824
                                        21.056
                                                       21.280
                                                                     827.34
                                         10.920
                                                       10.680
     12.0
               4.200
                        3.360
                                                                     878.41
                                         10.744
     13.6
               5.032
                        3.808
                                                       12.920
                                                                     835.50
                        3.888
     10.8
               4.968
                                          9.396
                                                        8.856
                                                                    1068.73
6
     16.2
                        4.860
                                         14.094
                                                       16.038
                                                                    1137.87
7
               6.156
      5.9
               2.006
                        1.593
                                          5.900
                                                        5.900
                                                                    1273.89
8
     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
                                         12.032
                                                       12.288
                                                                     803.11
13
     12.8
               4.608
                        4.352
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
                                                                     780.45
16
     17.8
               4.806
                        4.272
                                         13.706
                                                       15.130
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
                        2.870
      8.2
               1.886
                                          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

```
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
df.head()
          speeding alcohol
                              not distracted no previous
                                                             ins premium
0
    18.8
             7.332
                       5.640
                                       18.048
                                                    15.040
                                                                  784.55
    18.1
             7.421
                       4.525
                                       16.290
                                                    17.014
                                                                 1053.48
    18.6
                                                                  899.47
2
             6.510
                       5.208
                                       15.624
                                                    17.856
3
    22.4
                       5.824
                                       21.056
             4.032
                                                    21.280
                                                                  827.34
    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
                   ΑZ
3
       142.39
                   AR
4
       165.63
                   CA
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
                      51 non-null
                                       float64
     speeding
 2
                                       float64
     alcohol
                      51 non-null
3
                      51 non-null
                                       float64
     not distracted
4
                                       float64
     no previous
                      51 non-null
 5
     ins premium
                      51 non-null
                                       float64
     ins losses
                      51 non-null
                                       float64
6
7
     abbrev
                      51 non-null
                                       object
dtypes: float64(7), object(1)
memory usage: 3.3+ KB
df.shape
```

```
(51, 8)
```

### ###Handling null values

```
df.isnull().any()
total
                   False
speeding
                   False
                   False
alcohol
not distracted
                   False
                   False
no_previous
ins_premium
                   False
ins losses
                   False
abbrev
                   False
dtype: bool
df.isnull().sum()
total
                   0
speeding
                   0
                   0
alcohol
                   0
not_distracted
                   0
no_previous
ins_premium
                   0
ins losses
                   0
abbrev
                   0
dtype: int64
```

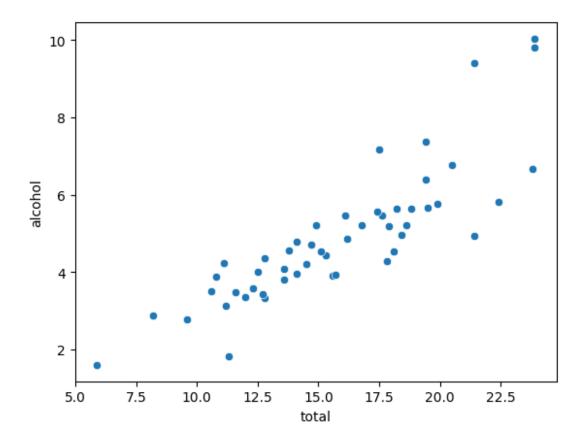
#### SEPARATION OF DEPENDENT AND INDEPENDENT VARIABLES

```
X=df.iloc[:,:1]
Χ
    total
0
     18.8
1
     18.1
2
     18.6
3
     22.4
4
     12.0
5
     13.6
6
     10.8
7
     16.2
8
     5.9
9
     17.9
10
     15.6
11
     17.5
     15.3
12
13
     12.8
14
     14.5
     15.7
15
```

```
17.8
16
     21.4
17
18
     20.5
19
     15.1
20
     12.5
21
     8.2
22
     14.1
23
     9.6
24
     17.6
25
     16.1
26
     21.4
     14.9
27
28
     14.7
29
     11.6
30
     11.2
31
     18.4
     12.3
32
     16.8
33
34
     23.9
35
     14.1
36
     19.9
37
     12.8
38
     18.2
39
     11.1
     23.9
40
41
     19.4
42
     19.5
43
     19.4
44
     11.3
45
     13.6
     12.7
46
47
     10.6
48
     23.8
49
     13.8
50
     17.4
```

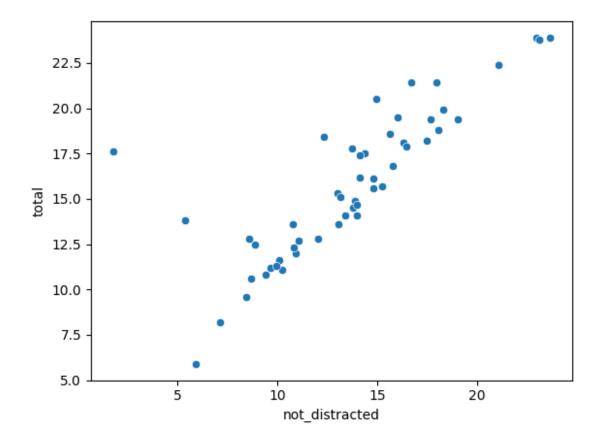
### Data visualisation

```
sns.scatterplot(x="total",y="alcohol",data=df)
<Axes: xlabel='total', ylabel='alcohol'>
```



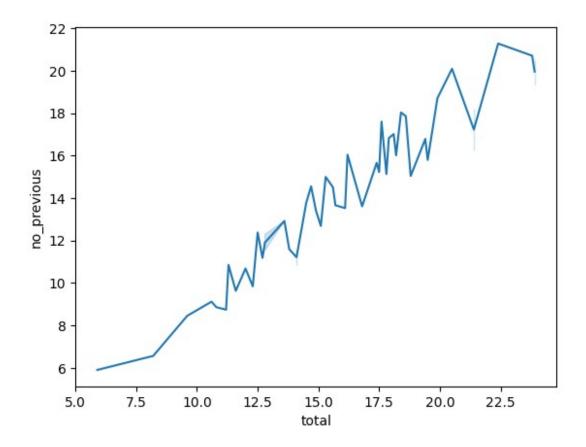
The graph infers that the total accidents that have taken place are directly proportional to the amount of alcohol consumed.

```
sns.scatterplot(y="total",x="not_distracted",data=df)
<Axes: xlabel='not_distracted', ylabel='total'>
```



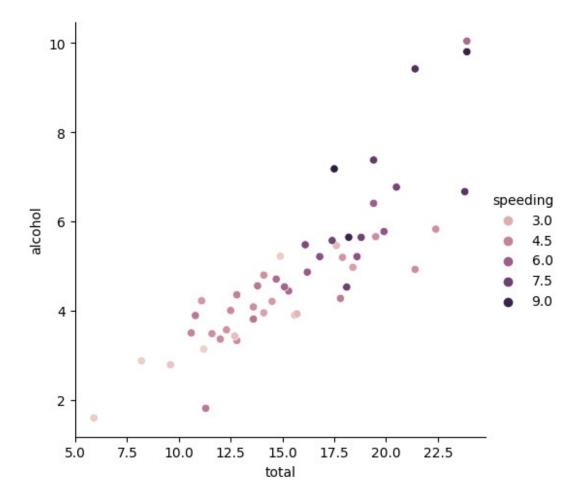
Here the graph infers that the total accidents that have taken place are directly proportional to the accidents that have taken place because of not being distracted.

```
sns.lineplot(x="total",y="no_previous",data=df)
<Axes: xlabel='total', ylabel='no_previous'>
```



Here the graph infers that the total accidents that have taken place are directly proportional to accidents that not taken place prior.

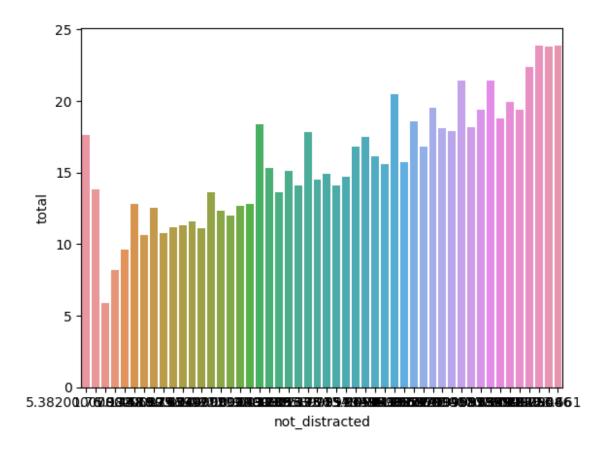
```
sns.relplot(x="total",y="alcohol",data=df,hue="speeding")
<seaborn.axisgrid.FacetGrid at 0x7e8085c19390>
```



Here the graph infers that the total accidents that have taken place due to consumption of alcohol are directly proportional to speed with which the car was driven

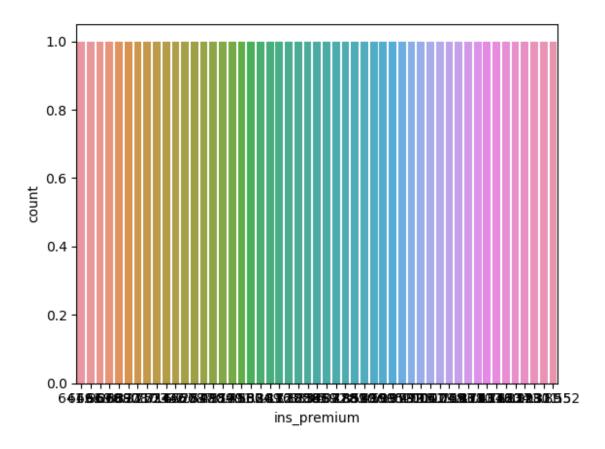
```
sns.barplot(x="not_distracted",y="total",data=df,ci=None)
<ipython-input-29-3c11a350536e>:1: FutureWarning:
The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

sns.barplot(x="not_distracted",y="total",data=df,ci=None)
<Axes: xlabel='not_distracted', ylabel='total'>
```



The graph infers that the total accidents that have taken place are directly proportional to the reason being that the driver was not distracted

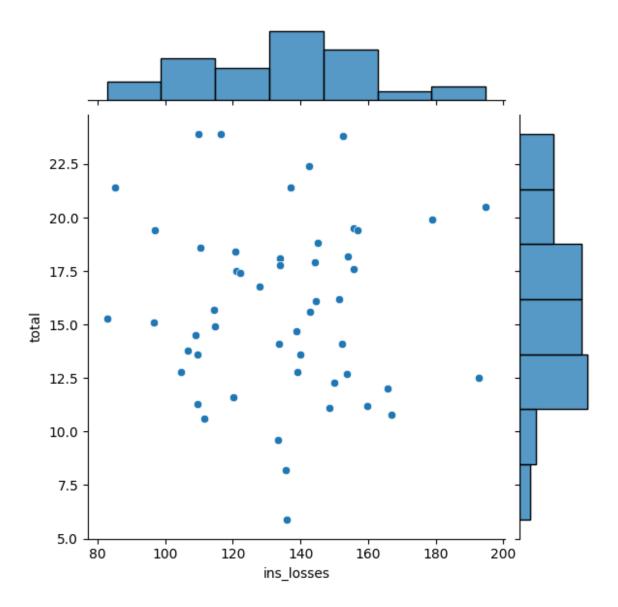
```
sns.countplot(x="ins_premium",data=df)
<Axes: xlabel='ins_premium', ylabel='count'>
```



## The graph gives us the count of total ins\_premium

sns.jointplot(x="ins\_losses",y="total",data=df)

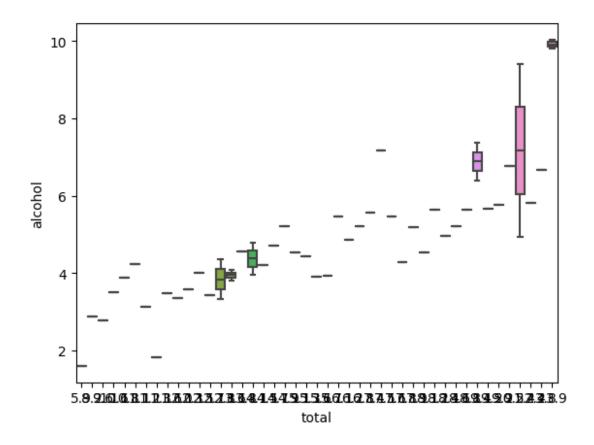
<seaborn.axisgrid.JointGrid at 0x7e8088bb9870>



# The above graph gives data plotted in two differnt types.

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

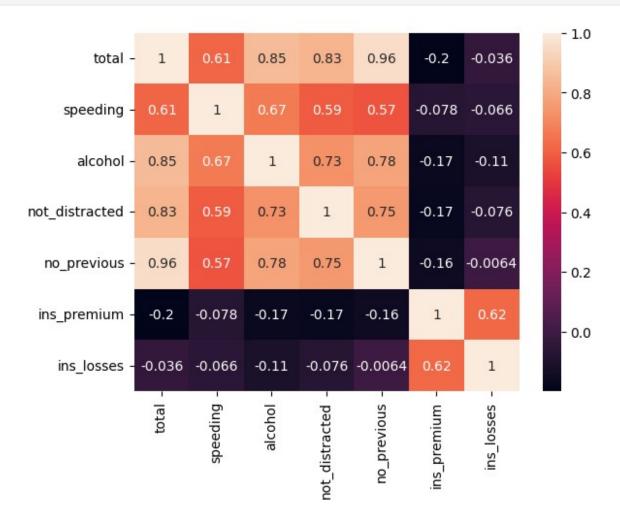
<Axes: xlabel='total', ylabel='alcohol'>



Here the graph infers that the total accidents that have taken place are directly proportional to the amount of alcohol consumed

```
c=df.corr()
<ipython-input-33-d9e4e726f1fe>: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.
  c=df.corr()
                   total
                           speeding
                                      alcohol not_distracted
no previous
total
                1.000000
                          0.611548
                                     0.852613
                                                      0.827560
0.956179
                0.611548 1.000000
                                     0.669719
                                                      0.588010
speeding
0.571976
alcohol
                0.852613
                          0.669719
                                     1.000000
                                                      0.732816
0.783520
not distracted
                0.827560 0.588010
                                     0.732816
                                                      1.000000
0.7\overline{4}7307
```

no_previous 1.000000	0.956179 0.	571976	0.783520	0.747307	'						
ins_premium 0.156895	-0.199702 -0.	077675	-0.170612	-0.174856	j -						
ins_losses 0.006359	-0.036011 -0.	065928	-0.112547	-0.075970	-						
total speeding alcohol not_distracted no_previous ins_premium ins_losses	ins_premium -0.199702 -0.077675 -0.170612 -0.174856 -0.156895 1.000000 0.623116	-0.06 -0.11 -0.07 -0.06	86011 55928 12547 75970								
<pre>sns.heatmap(c,annot=True)</pre>											
<axes:></axes:>											



The graph shows us the intensity of variables of the above mentioned dataset