DEVANSHI PATEL

assignment2

September 13, 2023

```
import numpy as np
[1]:
     import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
[2]: df = sns.load_dataset('car_crashes')
     df
[2]:
         total
                 speeding
                            alcohol
                                      not_distracted
                                                       no_previous
                                                                      ins_premium \
           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
                                                                           899.47
                                               15.624
                                                             17.856
     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
                    5.508
                              4.437
           15.3
                                               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
           17.8
                              4.272
     16
                    4.806
                                               13.706
                                                             15.130
                                                                           780.45
     17
          21.4
                    4.066
                              4.922
                                               16.692
                                                             16.264
                                                                           872.51
     18
          20.5
                              6.765
                                                                          1281.55
                    7.175
                                               14.965
                                                             20.090
     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
                                                                          1011.14
                                                              6.560
     22
           14.1
                    3.384
                              3.948
                                               13.395
                                                             10.857
                                                                          1110.61
     23
           9.6
                    2.208
                              2.784
                                                                           777.18
                                                8.448
                                                              8.448
          17.6
     24
                    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 | ${\tt abbrev}$ |
|----|------------|----------------|
| 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
                    ΜI
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
                    NY
        150.01
33
        127.82
                    NC
34
                    ND
        109.72
35
        133.52
                    OH
36
        178.86
                    OK
37
        104.61
                    OR
38
                    PA
        153.86
39
        148.58
                    RΙ
40
                    SC
        116.29
41
         96.87
                    SD
42
        155.57
                    TN
43
        156.83
                    TX
44
        109.48
                    UT
45
                    VT
        109.61
46
        153.72
                    VA
47
        111.62
                    WA
48
        152.56
                    WV
49
        106.62
                    WΙ
50
        122.04
                    WY
```

[3]: df.info() #information

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 51 entries, 0 to 50
Data columns (total 8 columns):

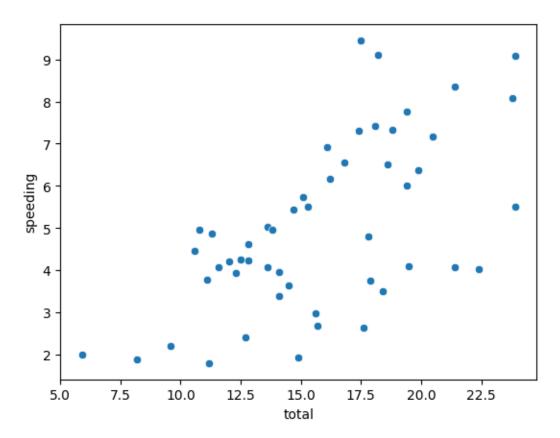
| # | Column | Non-Null Count | Dtype |
|---|-------------------------|----------------|---------|
| | | | |
| 0 | total | 51 non-null | float64 |
| 1 | speeding | 51 non-null | float64 |
| 2 | alcohol | 51 non-null | float64 |
| 3 | ${\tt not_distracted}$ | 51 non-null | float64 |
| 4 | no_previous | 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

```
[4]: sns.scatterplot(x="total",y="speeding",data=df)
```

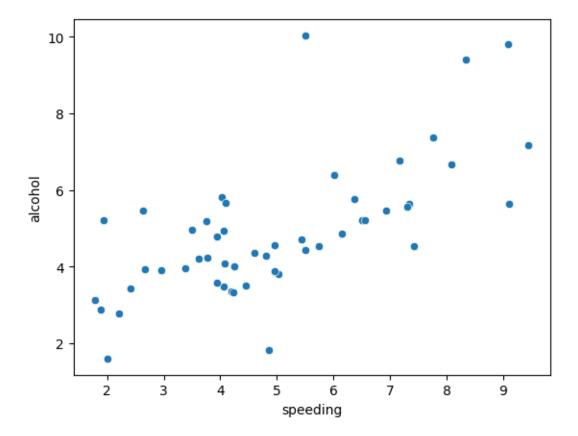
[4]: <Axes: xlabel='total', ylabel='speeding'>



Inference – From the above scatter graph we can see that it is positive weak correlation graph.

```
[5]: sns.scatterplot(x="speeding",y="alcohol",data=df)
```

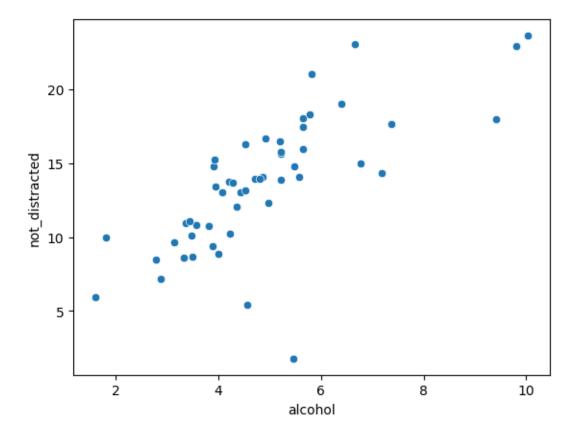
[5]: <Axes: xlabel='speeding', ylabel='alcohol'>



Inference – From the above scatter graph we can see that it is positive correlation graph.

```
[6]: sns.scatterplot(x="alcohol",y="not_distracted",data=df)
```

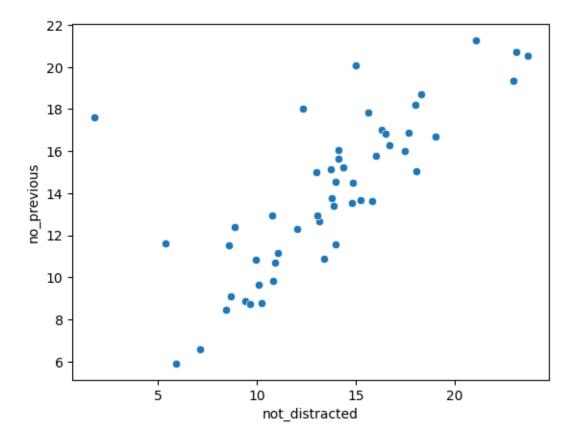
[6]: <Axes: xlabel='alcohol', ylabel='not_distracted'>



Inference – From the above scatter graph we can see that it is positive correlation graph.

```
[7]: sns.scatterplot(x="not_distracted",y="no_previous",data=df)
```

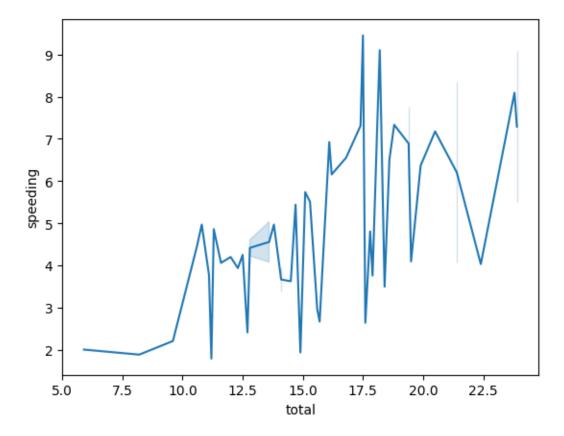
[7]: <Axes: xlabel='not_distracted', ylabel='no_previous'>



Inference – From the above scatter graph we can see that it is positive correlation graph.

```
[8]: sns.lineplot(x="total",y="speeding",data=df)
```

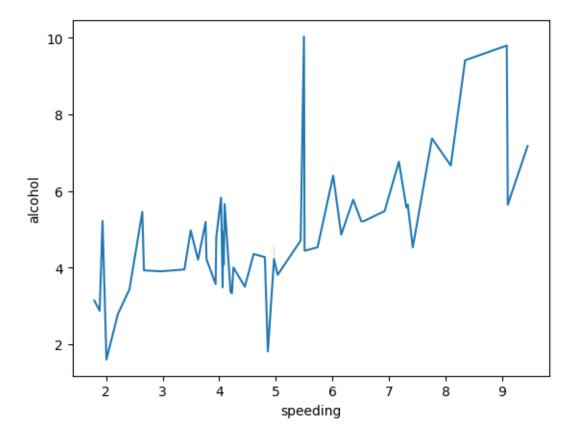
[8]: <Axes: xlabel='total', ylabel='speeding'>



Inference – From the above line graph we can see that there is no such particular relation between total and speeding values i.e., sometimes it increases and sometimes it decreases.

```
[9]: sns.lineplot(x="speeding",y="alcohol",data=df)
```

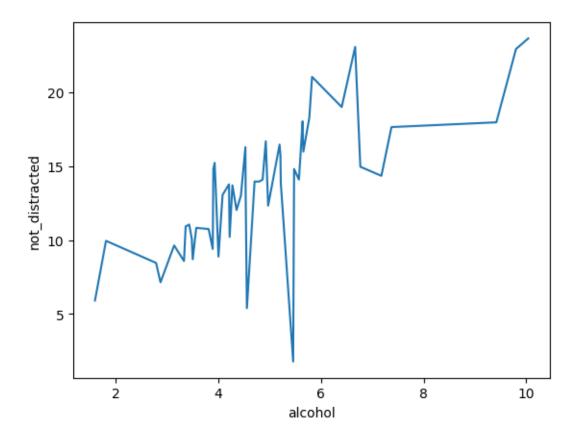
[9]: <Axes: xlabel='speeding', ylabel='alcohol'>



Inference – From the above line graph we can see that there is no such particular relation between speeding and alcohol values i.e., sometimes it increases and sometimes it decreases.

```
[10]: sns.lineplot(x="alcohol",y="not_distracted",data=df)
```

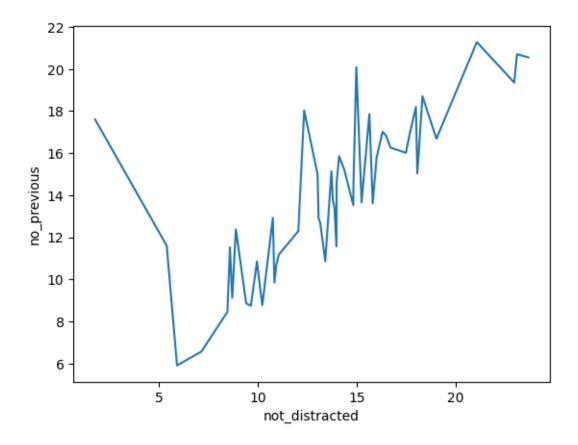
[10]: <Axes: xlabel='alcohol', ylabel='not_distracted'>



Inference – From the above line graph we can see that there is no such particular relation between alcohol and not distracted values i.e., sometimes it increases and sometimes it decreases but most of the time it increases.

```
[11]: sns.lineplot(x="not_distracted",y="no_previous",data=df)
```

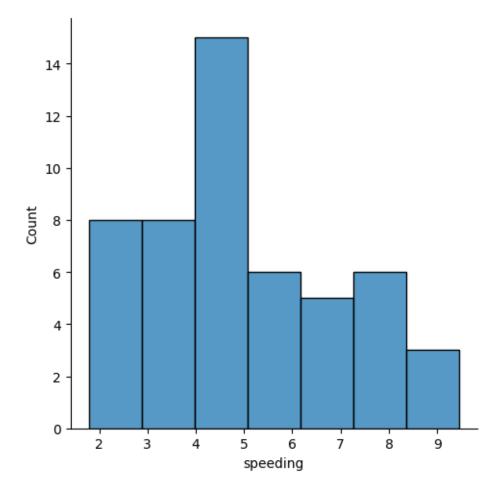
[11]: <Axes: xlabel='not_distracted', ylabel='no_previous'>



Inference – From the above line graph we can see that there is no such particular relation between not distracted and no previous values i.e., sometimes it increases and sometimes it decreases but at starting first the value drops then it's increasing gradually/

```
[21]: sns.displot(df["speeding"])
```

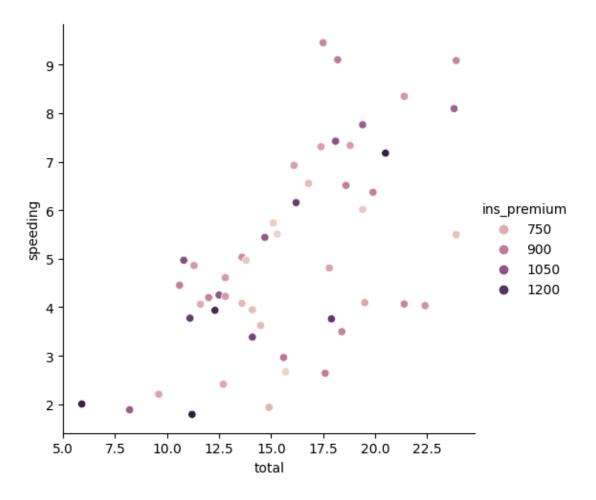
[21]: <seaborn.axisgrid.FacetGrid at 0x7ba72e052ad0>



Inference – Between 4 to 5 the count value is maximum and in different speeding values the average is around 6.

```
[13]: sns.relplot(x="total",y="speeding",data=df,hue="ins_premium")
```

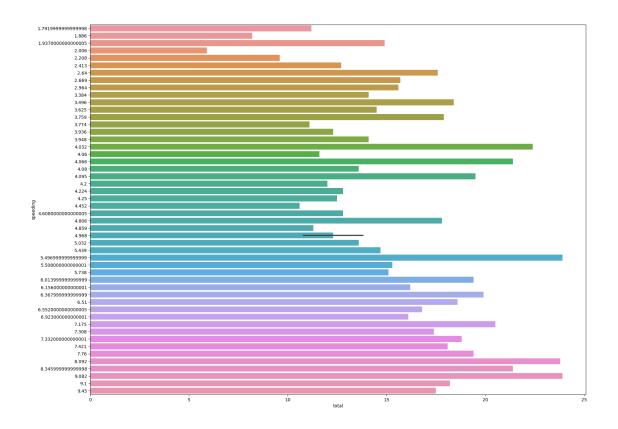
[13]: <seaborn.axisgrid.FacetGrid at 0x7ba72dd1fd30>



Inference – The above graph is a scatter plot with different colors which indicates different values.

```
[14]: plt.subplots(figsize=(20,15))
sns.barplot(data=df,x="total",y="speeding",orient='h')
```

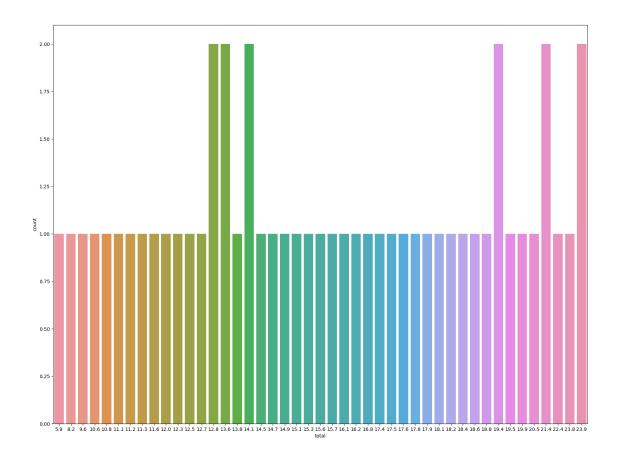
[14]: <Axes: xlabel='total', ylabel='speeding'>



Inference – The above graph is bar plot which is total vs speeding graph.

```
[15]: plt.subplots(figsize=(20,15))
sns.countplot(x="total",data=df)
```

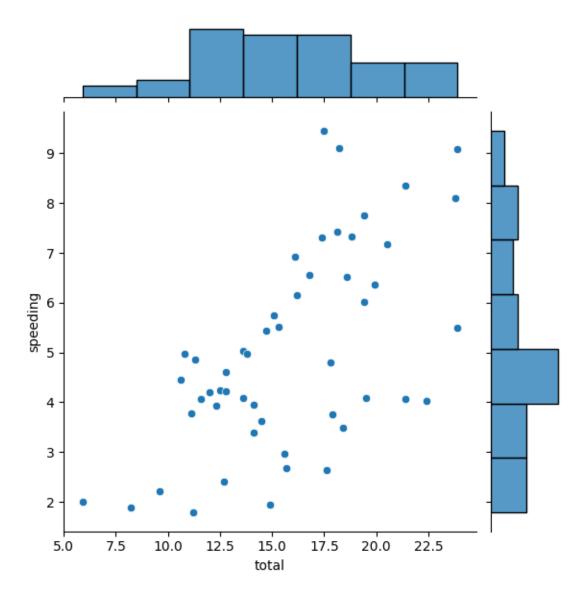
[15]: <Axes: xlabel='total', ylabel='count'>



Inference – The above count graph is of total vs count and the maximum count is around 2.

```
[16]: sns.jointplot(x="total",y="speeding",data=df)
```

[16]: <seaborn.axisgrid.JointGrid at 0x7ba72ba1a050>

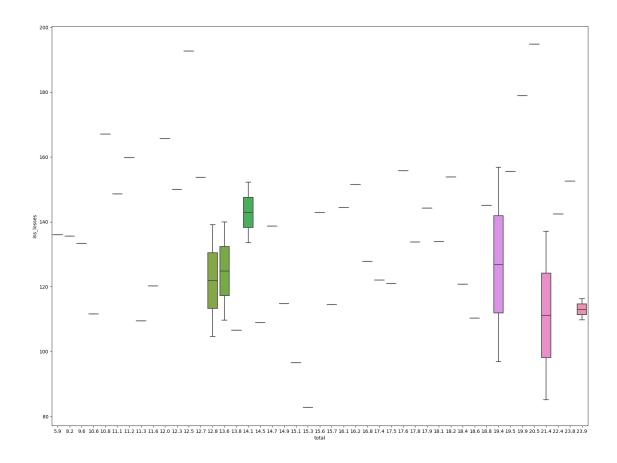


Inference – The above graph is joint graph which is combination of scatter plot and histogram or bar plot.

#Box Plot

```
[17]: plt.subplots(figsize=(20,15))
sns.boxplot(x="total",y="ins_losses",data=df)
```

[17]: <Axes: xlabel='total', ylabel='ins_losses'>



[18]: corr = df.corr()
corr

<ipython-input-18-4381f08f6434>: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.

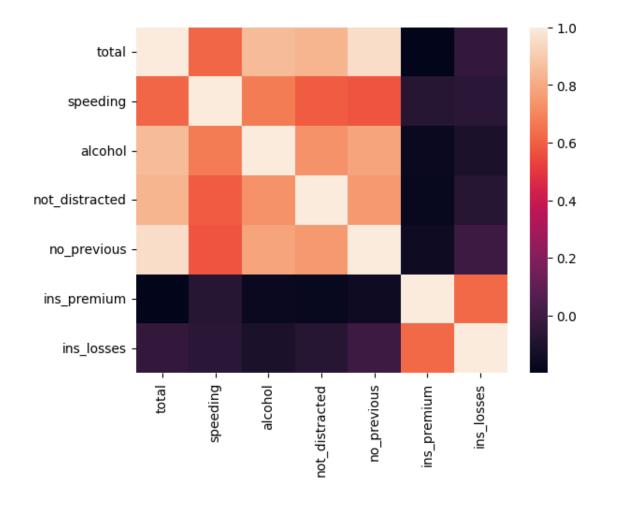
corr = df.corr()

[18]: 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 0.747307 no_previous 0.956179 0.571976 0.783520 0.747307 1.000000 -0.199702 -0.077675 -0.170612 ins_premium -0.174856 -0.156895 ins_losses -0.036011 -0.065928 -0.112547 -0.075970 -0.006359

ins_premium ins_losses total -0.199702 -0.036011 speeding -0.077675 -0.065928 alcohol-0.170612-0.112547not_distracted-0.174856-0.075970no_previous-0.156895-0.006359ins_premium1.0000000.623116ins_losses0.6231161.000000

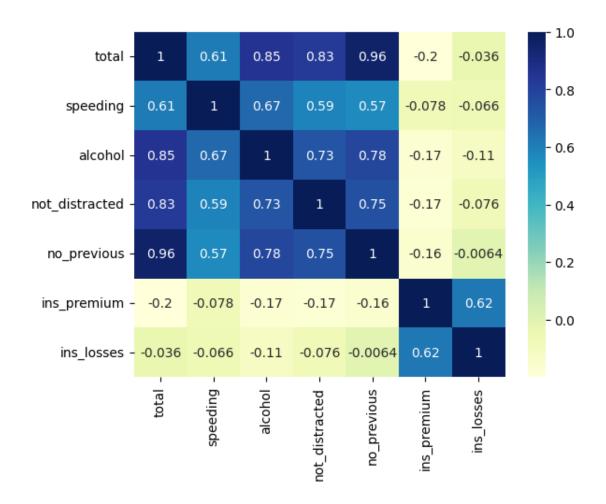
[19]: sns.heatmap(corr)

[19]: <Axes: >



[20]: sns.heatmap(corr,annot=True,cmap="YlGnBu")

[20]: <Axes: >



[20]: