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# Assignment2

### September 13, 2023

```
[1]: import numpy as np
     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 \
     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
           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
                2.006 1.593 5.900 5.900 1273.89
                                       16.826
                                                   1160.13
     9
          17.9 3.759 5.191 16.468
     10
          15.6 2.964 3.900 14.820
                                       14.508
                                                   913.15
          17.5 9.450 7.175 14.350
                                       15.225
                                                   861.18
     11
          15.3 5.508 4.437 13.005
     12
                                       14.994
                                                   641.96
     13
          12.8 4.608 4.352 12.032
                                       12.288
                                                   803.11
          14.5 3.625 4.205 13.775
     14
                                       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
          21.4 4.066 4.922 16.692 16.264 872.51 18 20.5 7.175
     17
          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
     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
          14.7 5.439 4.704 13.965
     28
                                       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
```

```
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  12.3 3.936 3.567 10.824 9.840 1234.31
32
33 16.8 6.552 5.208 15.792 13.608 708.24
   23.9 5.497 10.038 23.661 20.554 688.75
34
                            11.562 697.73
35
    14.1 3.948 4.794 13.959
    19.9 6.368 5.771 18.308
                            18.706
36
                                      881.51
   12.8 4.224 3.328 8.576 11.520 804.71
37
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
  11.3 4.859 1.808 9.944 10.848 809.38
44
  13.64.080 4.080 13.05612.920716.2012.72.413 3.429 11.04911.176768.95
45
46
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
```

|    | TII2_TO2262 | applev      |    |
|----|-------------|-------------|----|
| 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      | ${	t FL}$   |    |
| 10 | 142.80      | GA          |    |
| 11 | 120.92      | HI          |    |
| 12 | 82.75 II    | D 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 M     | E 20 192.70 | MD |
| 21 | 135.63      | MA          |    |
| 22 | 152.26      | MI          |    |
| 23 | 133.35      | MN          |    |
| 24 | 155.77      | MS          |    |

144.45 MO

25

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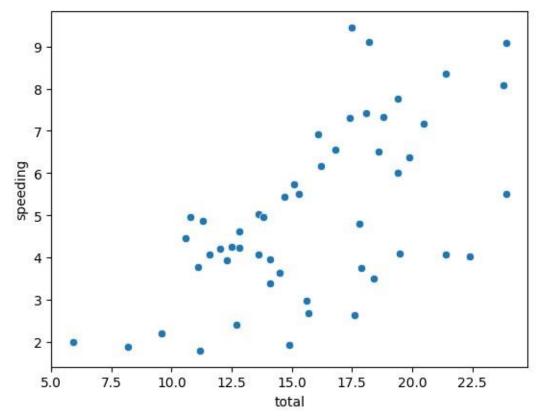
```
26
          85.15
                    МТ
    27
          114.82
                    ΝE
    28
          138.71
                    NV
    29
          120.21
                    NH
    30
          159.85
                    NJ
    31
          120.75
                    NM
    32
          150.01
                    NY
    33
          127.82
                    NC
    34
          109.72
                    ND
    35
          133.52
                    ОН
    36
          178.86
                    OK
    37
          104.61
                    OR
    38
          153.86
                    PA
    39
          148.58
                    RΙ
    40
          116.29
                    SC
    41
                    SD
          96.87
    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
                    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 not distracted 51 non-null float64
   4 no previous
                   51 non-null float64
      ins premium
                    51 non-null float64
   5
   6
     ins losses
                    51 non-null float64
                    51 non-null
       abbrev
                                  object
   dtypes: float64(7), object(1)
```

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

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

memory usage: 3.3+ KB

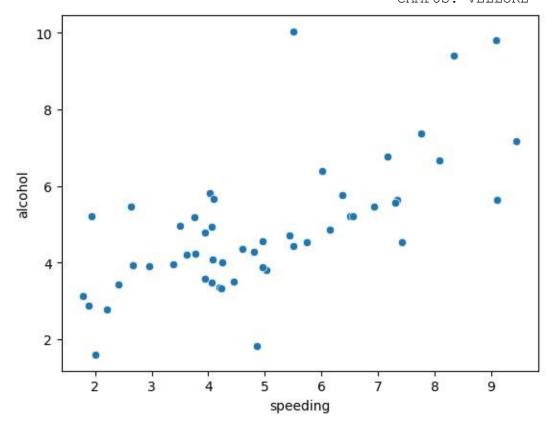
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Inference – From the above scatter graph we can see that it is positive weak correlation graph.

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

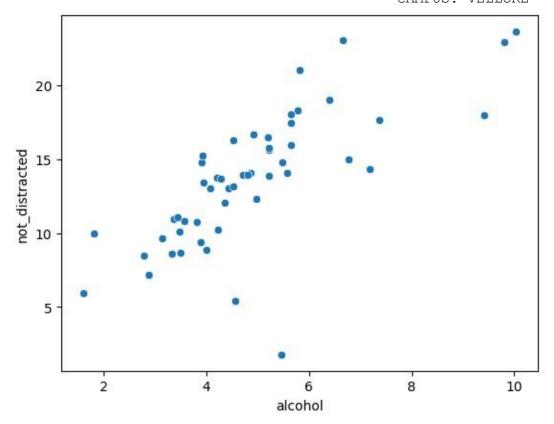
[5]: <Axes: xlabel='speeding', ylabel='alcohol'>



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'>

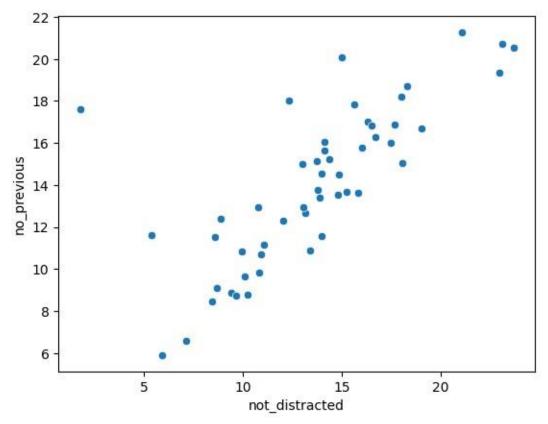


Inference – From the above scatter graph we can see that it is positive correlation graph.

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

[7]: <Axes: xlabel='not\_distracted', ylabel='no\_previous'>

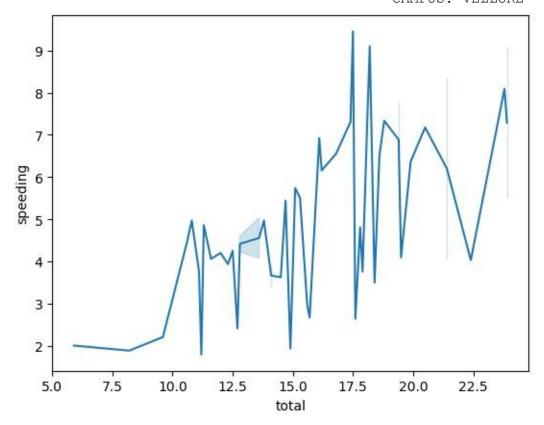
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Inference – From the above scatter graph we can see that it is positive correlation graph.

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

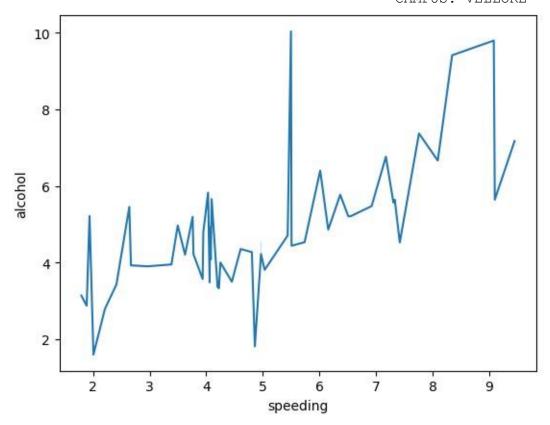
[8]: <Axes: xlabel='total', ylabel='speeding'>



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'>

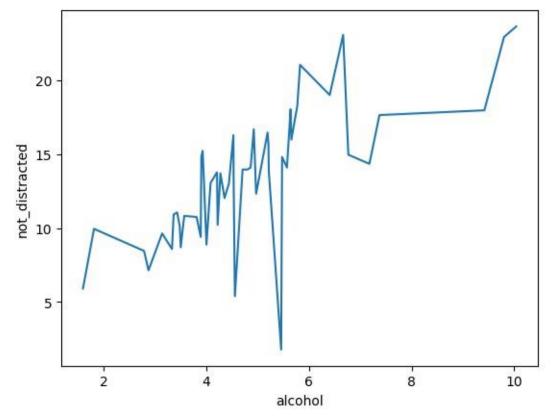


Inference – From the above line graph we can see that there is no such particular relation between speeding and alcohol values i.e., sometimes it increases and sometimes it decreases.

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

[10]: <Axes: xlabel='alcohol', ylabel='not distracted'>

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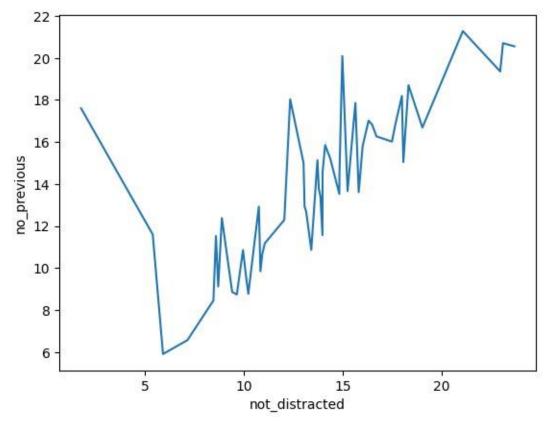


Inference – From the above line graph we can see that there is no such particular relation between alcohol and not distracted values i.e., sometimes it increases and sometimes it decreases but most of the time it increases.

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

[11]: <Axes: xlabel='not\_distracted', ylabel='no\_previous'>

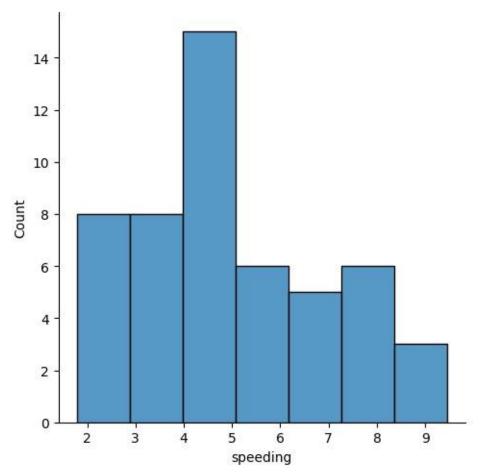
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Inference - From the above line graph we can see that there is no such particular relation between not distracted and no previous values i.e., sometimes it increases and sometimes it decreases but at starting first the value drops then it's increasing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

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

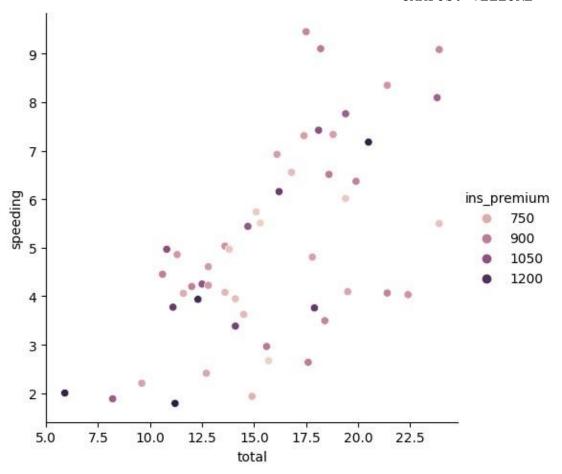
[21]: <seaborn.axisgrid.FacetGrid at 0x7ba72e052ad0>



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")
```

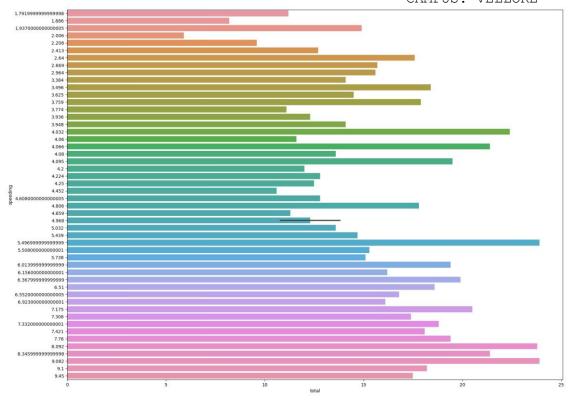
[13]: <seaborn.axisgrid.FacetGrid at 0x7ba72dd1fd30>



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')
```

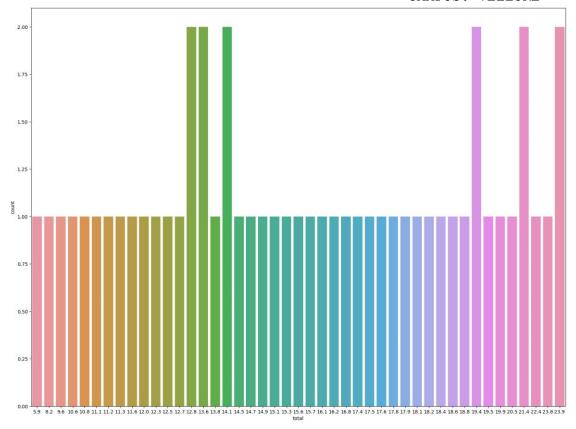
[14]: <Axes: xlabel='total', ylabel='speeding'>



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)
```

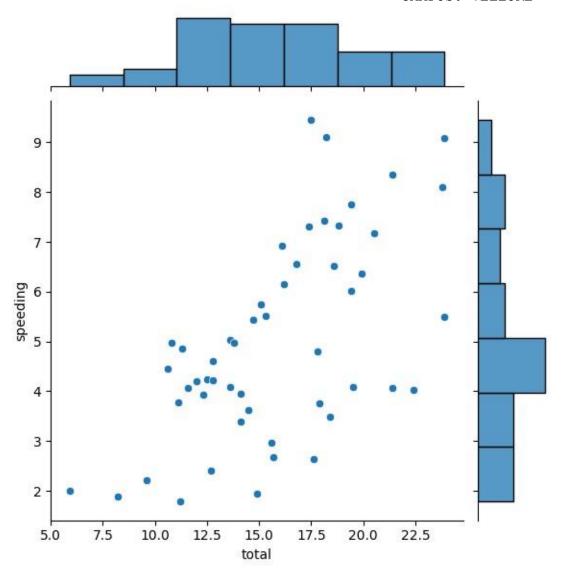
[15]: <Axes: xlabel='total', ylabel='count'>



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>



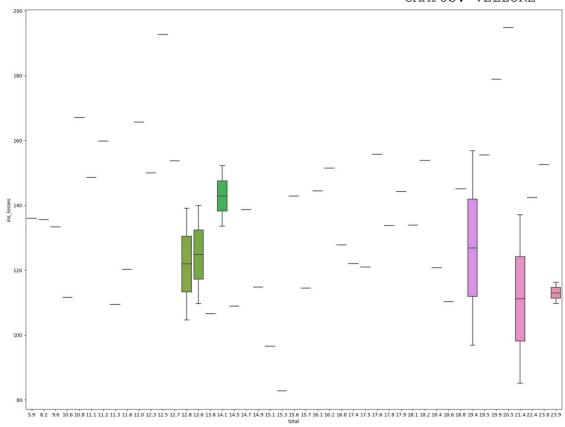
Inference – The above graph is joint graph which is combination of scatter plot and histogram or bar plot.

### **#Box Plot**

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

[17]: <Axes: xlabel='total', ylabel='ins\_losses'>

-0.174856 -0.156895



```
[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()

```
total speeding alcohol not distracted no previous
[18]:
    total
                  1.000000 0.611548 0.852613
                                                   0.827560
                                                              0.956179
    speeding
                  0.611548 1.000000 0.669719
                                                   0.588010
                                                              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
```

ins premium -0.199702 -0.077675 -0.170612

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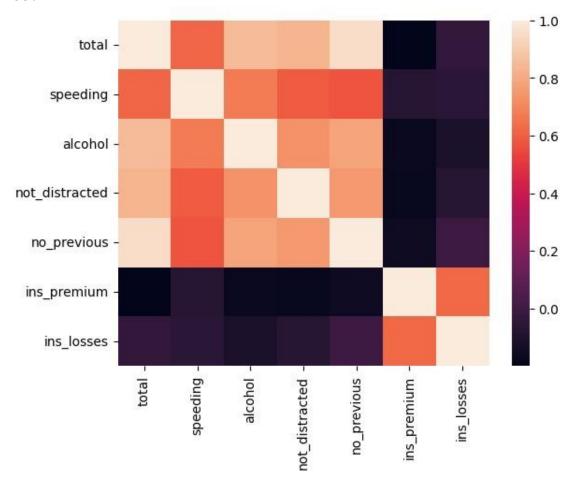
ins losses -0.036011 -0.065928 -0.112547 -0.075970 -0.006359

0.623116

1.000000

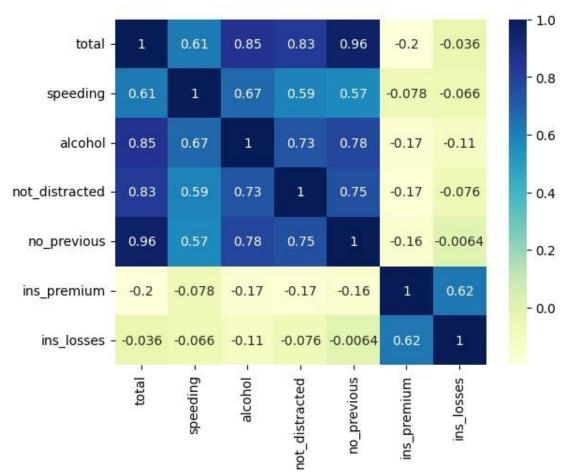
ins\_losses
[19]: sns.heatmap(corr)

### [19]: <Axes: >



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

### [20]: <Axes: >



## [20]: