21bit0481-asst-2-1

September 13, 2023

```
[34]:
      import seaborn as sns
      import matplotlib.pyplot as plt
 [3]: df=sns.load_dataset('car_crashes')
 [3]:
           total
                  speeding
                              alcohol
                                        not_distracted
                                                         no_previous
                                                                        ins_premium
                      7.332
      0
            18.8
                                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
                                3.360
            12.0
                      4.200
                                                 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
                                3.900
            15.6
                      2.964
                                                 14.820
                                                               14.508
                                                                              913.15
      11
            17.5
                                7.175
                                                               15.225
                      9.450
                                                 14.350
                                                                              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
                                4.205
      14
            14.5
                      3.625
                                                 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
                                                                              780.45
                                                               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
             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
                                                  8.448
                                                                              777.18
                                                                8.448
      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
                                9.416
                      8.346
                                                 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
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	MI
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
        150.01
                    NY
33
        127.82
                    NC
34
        109.72
                    ND
35
        133.52
                    OH
36
        178.86
                    OK
37
        104.61
                    OR
38
        153.86
                    PA
39
        148.58
                    RI
40
                    SC
        116.29
41
         96.87
                    SD
42
                    TN
        155.57
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
                    WY
        122.04
```

[4]: 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	${\tt not_distracted}$	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

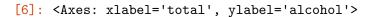
dtypes: float64(7), object(1)

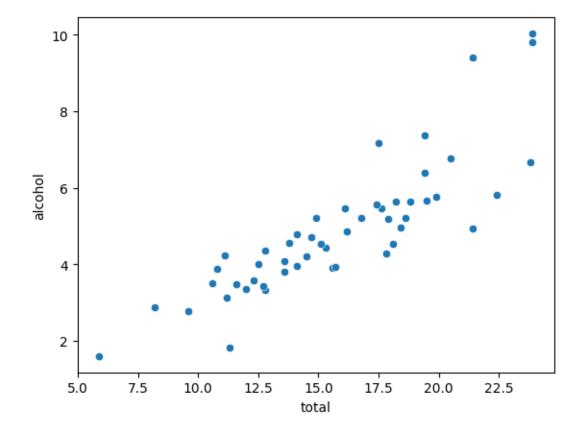
memory usage: 3.3+ KB

[5]: df.head(5)

```
[5]:
        total
                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
     1
     2
         18.6
                   6.510
                            5.208
                                             15.624
                                                           17.856
                                                                         899.47
                   4.032
     3
         22.4
                            5.824
                                             21.056
                                                           21.280
                                                                        827.34
         12.0
                   4.200
                            3.360
                                             10.920
                                                           10.680
                                                                        878.41
        ins_losses abbrev
     0
            145.08
                        AL
            133.93
                        AK
     1
     2
                        ΑZ
            110.35
     3
            142.39
                        AR
     4
            165.63
                        CA
```

[6]: sns.scatterplot(x='total',y='alcohol',data=df)





[7]: #Inference: from the plot we can say that the total car crashes increases with $_{\!\!\!\!\perp}$ $_{\!\!\!\!\perp}$ the increase in alcohol consumption.

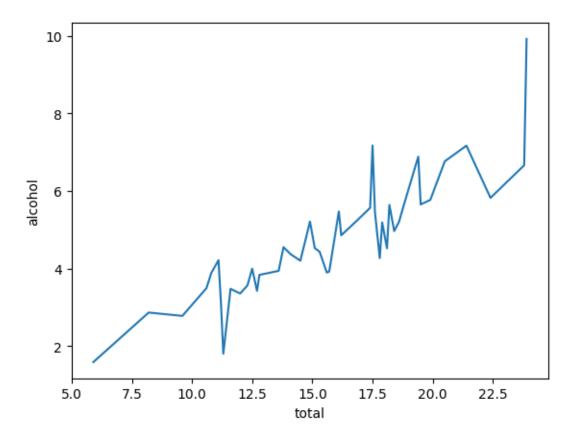
[8]: sns.lineplot(x='total',y='alcohol',data=df,ci=None)

<ipython-input-8-9720dd68e387>:1: FutureWarning:

The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

sns.lineplot(x='total',y='alcohol',data=df,ci=None)

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



[9]: #Inference: from the plot we can say that the total car crashes increases with the increse in alcohol consumption.

[10]: sns.distplot(df['alcohol'])

<ipython-input-10-570de8ff0310>:1: UserWarning:

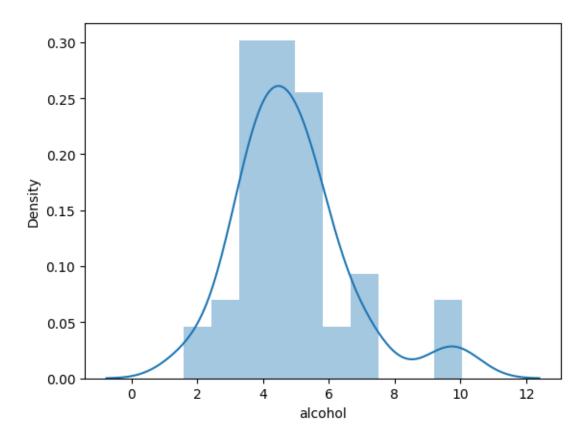
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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['alcohol'])

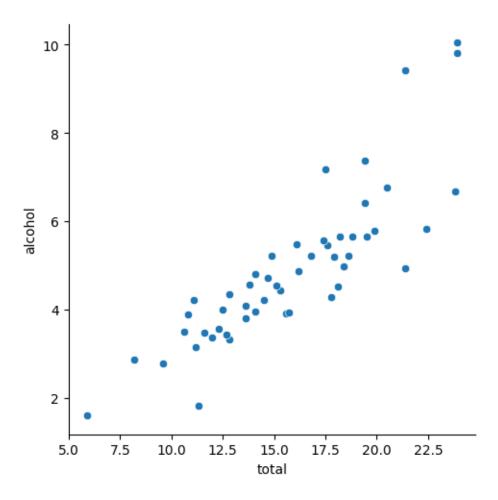
[10]: <Axes: xlabel='alcohol', ylabel='Density'>



[11]: #Inference: from the plot we can say that the maximum 0.30 leads to the mode of $_{\square}$ $_{\hookrightarrow}$ the distribution.

[12]: sns.relplot(x="total",y="alcohol",data=df)

[12]: <seaborn.axisgrid.FacetGrid at 0x7bdeec54ceb0>



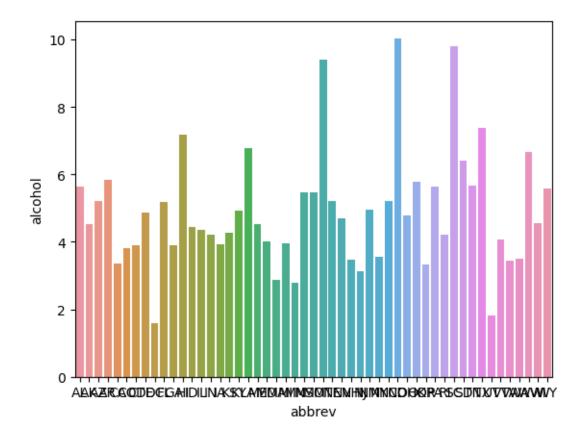
```
[13]: sns.barplot(data=df,x="abbrev",y="alcohol",ci=None)

<ipython-input-13-455bd05d1ba9>:1: FutureWarning:

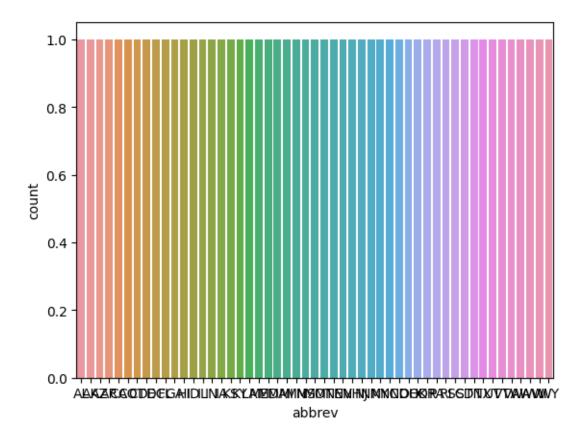
The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

sns.barplot(data=df,x="abbrev",y="alcohol",ci=None)
```

[13]: <Axes: xlabel='abbrev', ylabel='alcohol'>



- [14]: #inference: from the plot to compare the average percentage of alcohol-related \neg crashes across different states, represented by their abbreviations on the \neg x-axis.
- [15]: sns.countplot(data=df,x="abbrev")
- [15]: <Axes: xlabel='abbrev', ylabel='count'>

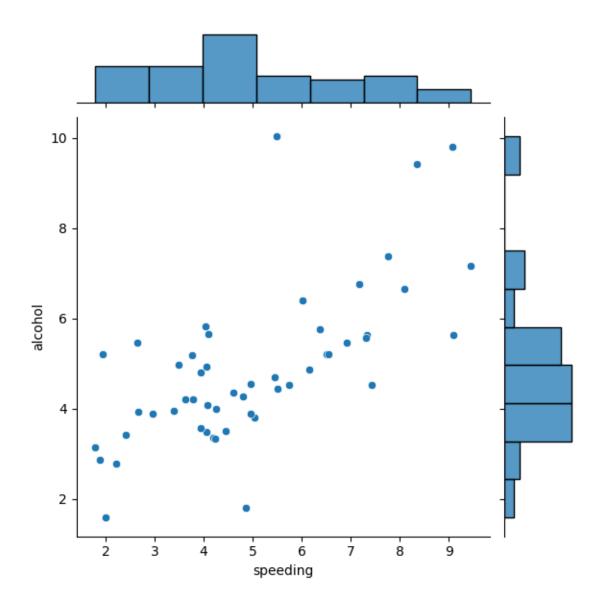


```
[16]: #inference: from the plot we can say that there is no variations in the 

→frequency distribution
```

[17]: sns.jointplot(data=df,x="speeding",y="alcohol")

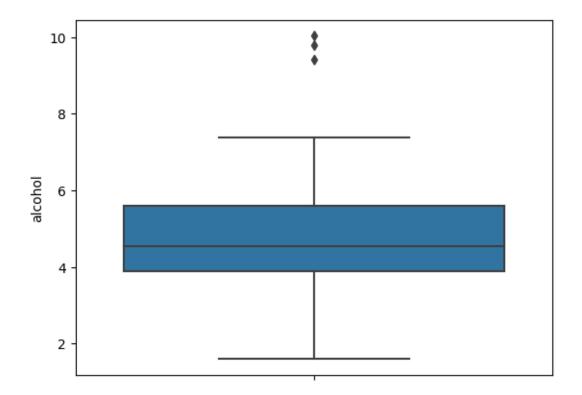
[17]: <seaborn.axisgrid.JointGrid at 0x7bdea564ae00>



```
[18]: #inference: from the plot we can say the relation between speeding and alcoholus in car crashes
```

[19]: sns.boxplot(y="alcohol",data=df)

[19]: <Axes: ylabel='alcohol'>



```
[20]: """inference: from the plot we can see how it is representing:
-The interquartile range (IQR).
-The bottom edge of the box represents the first quartile (Q1)
-The top edge of the box represents the third quartile (Q3)"""
```

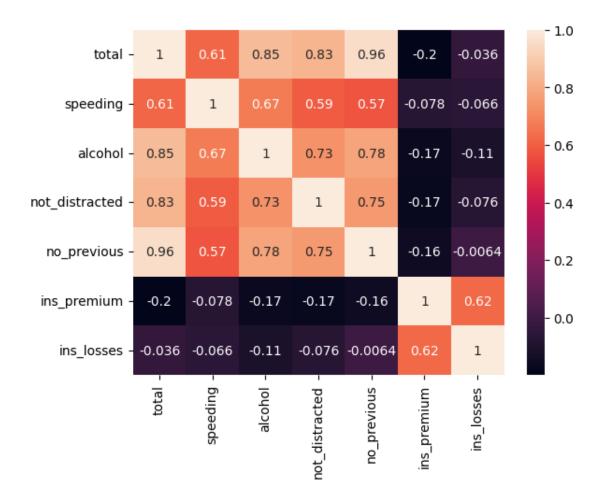
[20]: 'inference: from the plot we can see how it is representing:\n-The interquartile range (IQR).\n-The bottom edge of the box represents the first quartile (Q1)\n-The top edge of the box represents the third quartile (Q3)'

```
[21]: corr=df.corr() corr
```

<ipython-input-21-7d5195e2bf4d>: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()

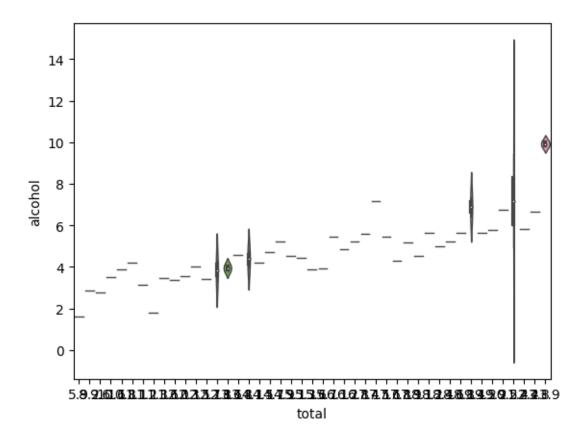
[21]: total speeding alcohol not_distracted no_previous \ total 1.000000 0.611548 0.852613 0.827560 0.956179 speeding 0.611548 1.000000 0.669719 0.588010 0.571976 alcohol 0.852613 0.669719 1.000000 0.732816 0.783520 not_distracted 0.827560 0.588010 0.732816 1.000000 0.747307

```
no_previous
                     0.956179 0.571976 0.783520
                                                         0.747307
                                                                      1.000000
     ins_premium
                    -0.199702 -0.077675 -0.170612
                                                                     -0.156895
                                                        -0.174856
     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
                                   -0.112547
     not_distracted
                       -0.174856
                                   -0.075970
     no_previous
                       -0.156895
                                   -0.006359
     ins_premium
                        1.000000
                                    0.623116
     ins_losses
                        0.623116
                                    1.000000
[22]: """#inference:
     >0.5 - highly correlated
      <0.5 - less correlated"""
[22]: '#inference:\n>0.5 - highly correlated\n<0.5 - less correlated'
[23]: sns.heatmap(corr,annot=True)
[23]: <Axes: >
```



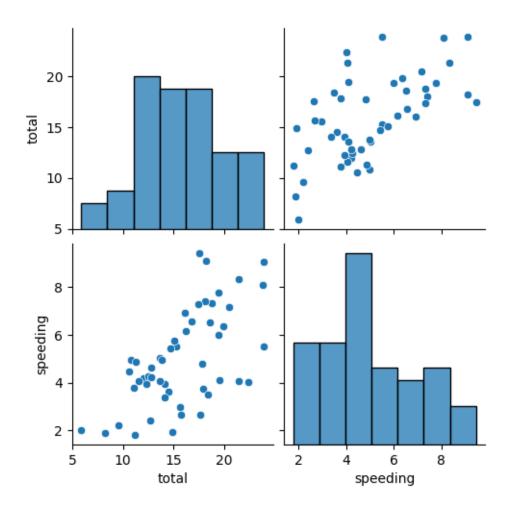
[24]: sns.violinplot(data=df,x='total',y='alcohol',inner='box',width=0.8,linewidth=1)

[24]: <Axes: xlabel='total', ylabel='alcohol'>



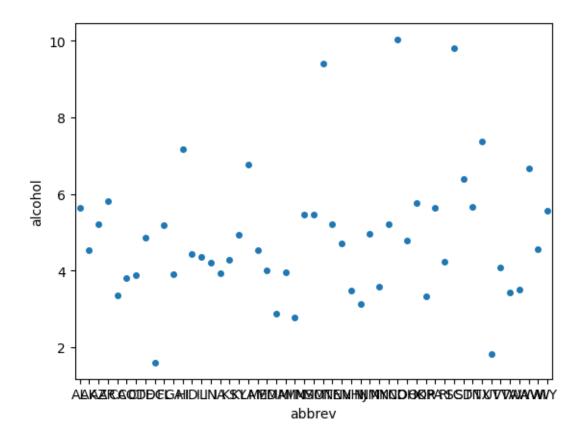
```
[28]: sns.pairplot(data=df, vars=["total", "speeding"])
```

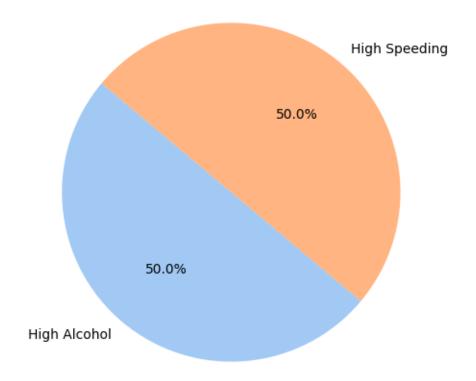
[28]: <seaborn.axisgrid.PairGrid at 0x7bdea1d5afe0>



```
[29]: #inference: Plotting and comaparing numerical variables total and speeding
[30]: sns.swarmplot(x="abbrev", y="alcohol", data=df)
```

[30]: <Axes: xlabel='abbrev', ylabel='alcohol'>





[]: #inference: from the plot we can say that the alcohol and speeding contribute_ to half of the total