assignment2

September 14, 2023

```
[1]:
     import seaborn as sns
    print(sns.get_dataset_names())
     ['anagrams', 'anscombe', 'attention', 'brain_networks', 'car_crashes',
     'diamonds', 'dots', 'dowjones', 'exercise', 'flights', 'fmri', 'geyser', 'glue',
     'healthexp', 'iris', 'mpg', 'penguins', 'planets', 'seaice', 'taxis', 'tips',
     'titanic']
[3]: df=sns.load_dataset('car_crashes')
[4]:
    df
[4]:
         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
                              3.808
           13.6
                    5.032
                                               10.744
                                                             12.920
                                                                           835.50
     6
           10.8
                    4.968
                              3.888
                                                9.396
                                                                          1068.73
                                                              8.856
     7
           16.2
                              4.860
                    6.156
                                               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
                                                             14.508
                                               14.820
                                                                           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
                              3.925
                                               15.229
           15.7
                    2.669
                                                             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
                              4.000
                    4.250
                                                8.875
                                                             12.375
                                                                          1048.78
     21
           8.2
                    1.886
                              2.870
                                                7.134
                                                              6.560
                                                                          1011.14
     22
                    3.384
                              3.948
                                               13.395
                                                             10.857
           14.1
                                                                          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		4.060		10.092	9.628	746.54
	11.6		3.480			
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

```
LA
18
        194.78
19
         96.57
                    ME
20
        192.70
                    MD
21
        135.63
                    MA
22
        152.26
                    ΜI
23
        133.35
                    MN
24
                    MS
        155.77
25
        144.45
                    MO
26
         85.15
                    MT
27
        114.82
                    NE
                    NV
28
        138.71
29
        120.21
                    NH
30
        159.85
                    NJ
31
        120.75
                    NM
32
        150.01
                    NY
33
                    NC
        127.82
34
        109.72
                    ND
35
        133.52
                    OH
36
        178.86
                    OK
37
                    OR
        104.61
38
        153.86
                    PA
39
                    RΙ
        148.58
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
                    WI
50
        122.04
                    WY
```

[5]: 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

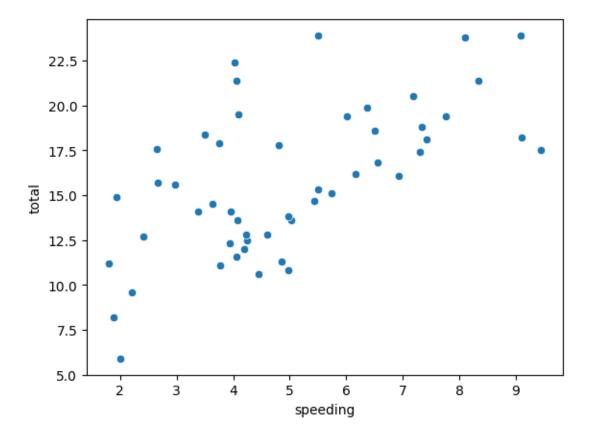
```
7
         abbrev
                           51 non-null
                                            object
    dtypes: float64(7), object(1)
    memory usage: 3.3+ KB
[6]: df.head()
[6]:
        total
               speeding alcohol not_distracted no_previous
                                                                   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
                            5.208
     2
                   6.510
                                             15.624
                                                          17.856
                                                                        899.47
     3
         22.4
                   4.032
                            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
     1
            133.93
                        AK
     2
            110.35
                        AZ
     3
            142.39
                        AR
     4
            165.63
                        CA
[7]: df.tail()
[7]:
                                     not_distracted no_previous
                                                                    ins_premium \
         total
                speeding
                           alcohol
                                                                         768.95
                    2.413
                             3.429
                                              11.049
     46
          12.7
                                                           11.176
     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
     46
             153.72
                         ۷A
             111.62
     47
                         WA
             152.56
                         WV
     48
     49
             106.62
                         WI
     50
             122.04
                         WY
[8]:
     df.shape
[8]: (51, 8)
[9]: df.isnull().sum()
[9]: total
                        0
     speeding
                        0
     alcohol
                        0
     not_distracted
                        0
```

float64

51 non-null

ins_losses

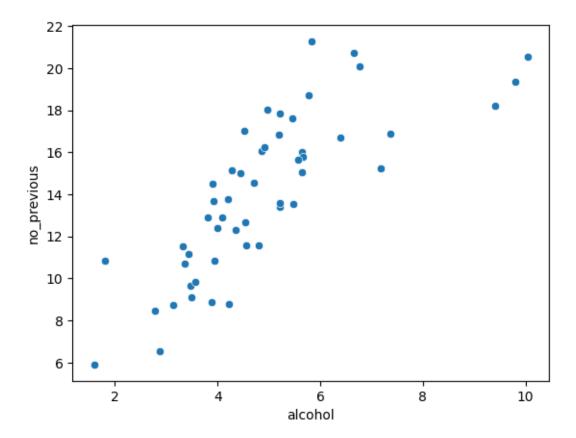
```
no_previous
                         0
                         0
      ins_premium
                         0
      ins_losses
                         0
      abbrev
      dtype: int64
[10]: df.describe()
[10]:
                          speeding
                                                not_distracted no_previous
                  total
                                       alcohol
      count
             51.000000
                         51.000000
                                    51.000000
                                                      51.000000
                                                                   51.000000
             15.790196
                          4.998196
                                      4.886784
                                                                    14.004882
      mean
                                                      13.573176
      std
              4.122002
                          2.017747
                                      1.729133
                                                       4.508977
                                                                     3.764672
      min
              5.900000
                          1.792000
                                      1.593000
                                                       1.760000
                                                                    5.900000
                          3.766500
      25%
             12.750000
                                      3.894000
                                                      10.478000
                                                                   11.348000
      50%
             15.600000
                          4.608000
                                      4.554000
                                                      13.857000
                                                                   13.775000
      75%
             18.500000
                          6.439000
                                      5.604000
                                                      16.140000
                                                                   16.755000
             23.900000
                          9.450000
                                     10.038000
                                                      23.661000
      max
                                                                   21.280000
                           ins_losses
             ins_premium
      count
               51.000000
                            51.000000
      mean
              886.957647
                           134.493137
      std
              178.296285
                            24.835922
      min
              641.960000
                            82.750000
      25%
              768.430000
                           114.645000
      50%
              858.970000
                           136.050000
      75%
             1007.945000
                           151.870000
             1301.520000
                           194.780000
      max
      sns.scatterplot(x='speeding',y='total',data=df)
[13]:
```



inference: increase in Percentage Of Drivers Involved In Fatal Collisions Who Were Speeding increases the Number of drivers involved in fatal collisions per billion miles

```
[55]: sns.scatterplot(x='alcohol',y='no_previous',data=df)
```

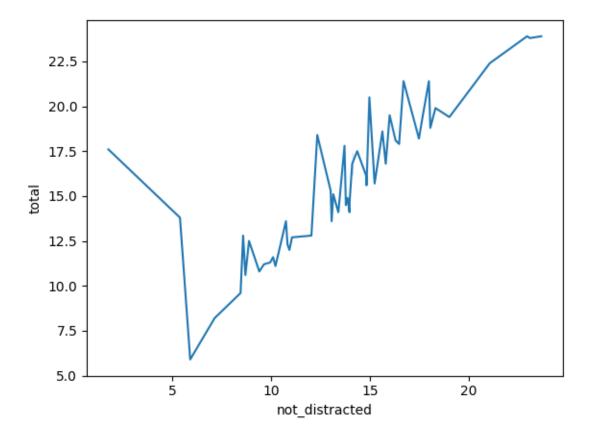
[55]: <Axes: xlabel='alcohol', ylabel='no_previous'>



inference: increase in Percentage Of Drivers Involved In Fatal Collisions Who Were Alcohol-Impaired increases the Percentage Of Drivers Involved In Fatal Collisions Who Had Not Been Involved In Any Previous Accidents

```
[20]: sns.lineplot(x='not_distracted',y='total',data=df)
```

[20]: <Axes: xlabel='not_distracted', ylabel='total'>



inference:not_distracted and total are not propotional.

[24]: sns.distplot(df['ins_losses'])

<ipython-input-24-dd3f37db2abe>:1: UserWarning:

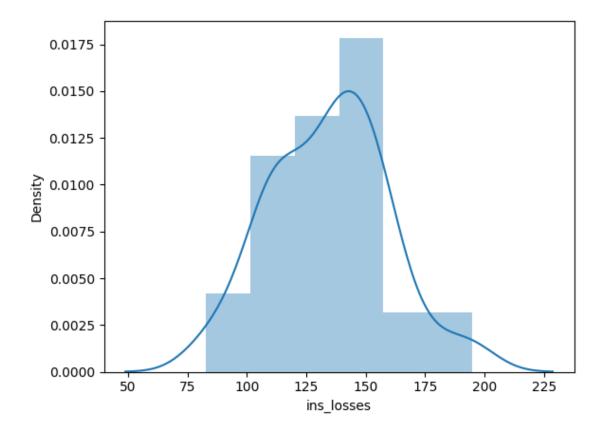
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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_losses'])

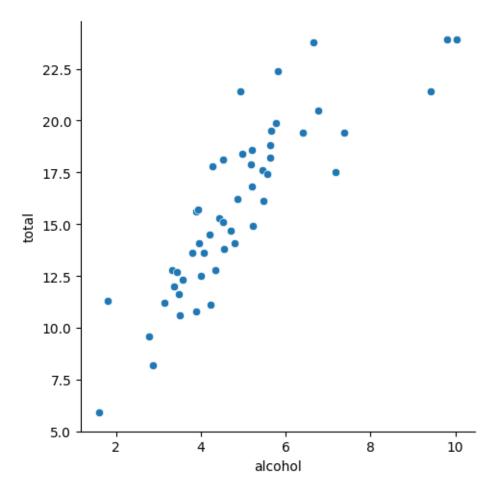
[24]: <Axes: xlabel='ins_losses', ylabel='Density'>



inference:distribution plot of insurnce loses

```
[29]: sns.relplot(y='total',x='alcohol',data=df)
```

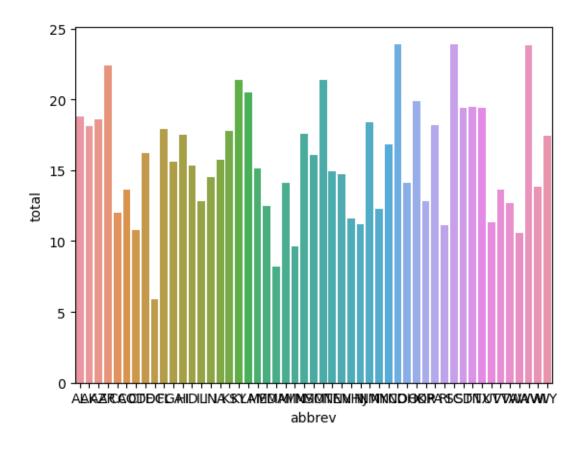
[29]: <seaborn.axisgrid.FacetGrid at 0x7f99d232bf10>



inference: increase in Percentage Of Drivers Involved In Fatal Collisions Who Were Alcohol-Impaired increases the Number of drivers involved in fatal collisions per billion miles

```
[32]: sns.barplot(x='abbrev',y='total',data=df)
```

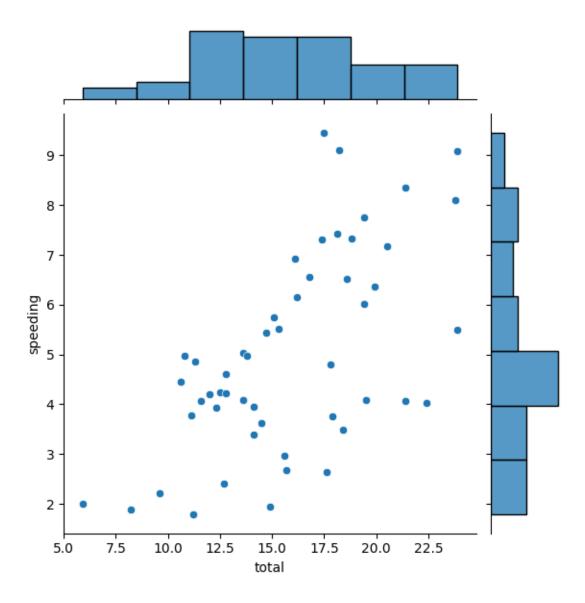
[32]: <Axes: xlabel='abbrev', ylabel='total'>



this the bar plot of different states of the USA and the total number of accidents.wv,sc,nd has highest accidents.

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

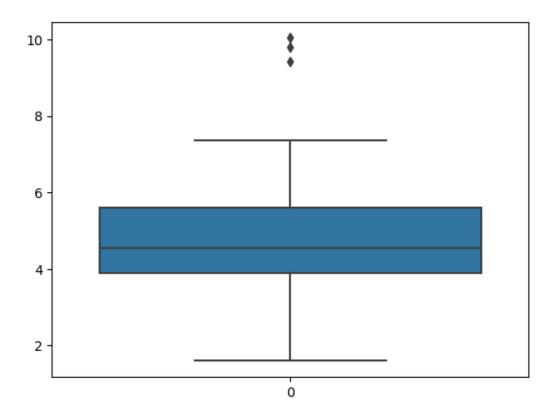
[37]: <seaborn.axisgrid.JointGrid at 0x7f99cfee8850>



joinplot between speeding and total number of accidents

[47]: sns.boxplot(df.alcohol)

[47]: <Axes: >



alcohol column has outliers

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

<ipython-input-52-8f785e6b066e>: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()#>0.5 highly correlated <0.5 less correlated</pre>

[52]:		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	-0.174856	-0.156895	
	ins_losses	-0.036011	-0.065928	-0.112547	-0.075970	-0.006359	
		ins_premi	ium ins_lo	sses			
	total	-0.1997	702 -0.03	36011			
	speeding	-0.0776	675 -0.06	55928			

 alcohol
 -0.170612
 -0.112547

 not_distracted
 -0.174856
 -0.075970

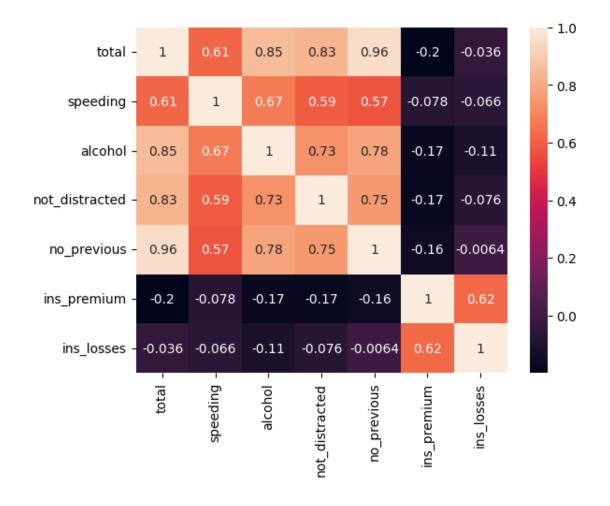
 no_previous
 -0.156895
 -0.006359

 ins_premium
 1.000000
 0.623116

 ins_losses
 0.623116
 1.000000

[53]: sns.heatmap(corr,annot=True)

[53]: <Axes: >



more than 0.5 are highly correlated, less than 0.5 less correlated

[]: