# ai-ml-assignment-2

## September 14, 2023

### Tarun Sreenivas 21BCE2106

```
[1]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
[]:|
     ak = sns.load_dataset('car_crashes')
     ak
[]:
                 speeding
                            alcohol
                                      not_distracted
                                                      no_previous
                                                                      ins_premium \
         total
           18.8
                    7.332
                              5.640
                                               18.048
                                                             15.040
                                                                            784.55
           18.1
                                               16.290
     1
                    7.421
                              4.525
                                                             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
           13.6
                    5.032
                              3.808
                                               10.744
                                                             12.920
                                                                            835.50
     6
           10.8
                    4.968
                              3.888
                                                9.396
                                                                           1068.73
                                                              8.856
     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
                              3.900
                    2.964
                                               14.820
                                                             14.508
                                                                            913.15
                    9.450
                              7.175
     11
           17.5
                                               14.350
                                                             15.225
                                                                            861.18
     12
           15.3
                    5.508
                              4.437
                                               13.005
                                                             14.994
                                                                            641.96
     13
                    4.608
                              4.352
           12.8
                                               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
                              4.272
     16
           17.8
                    4.806
                                               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
                    4.250
                              4.000
                                                8.875
                                                             12.375
                                                                           1048.78
     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
                              5.474
                    6.923
                                               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	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	116.29	SC
41	96.87	SD
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

# []: ak.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

```
[]: ak.head()
[]:
        total
               speeding alcohol not_distracted no_previous
                                                                   ins_premium \
     0
         18.8
                  7.332
                            5.640
                                            18.048
                                                          15.040
                                                                        784.55
                  7.421
                            4.525
     1
         18.1
                                            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
                                                                        827.34
                                                          21.280
                   4.200
     4
         12.0
                            3.360
                                            10.920
                                                          10.680
                                                                        878.41
        ins_losses abbrev
     0
            145.08
                        AL
     1
            133.93
                        ΑK
     2
            110.35
                        AZ
     3
                        AR
            142.39
     4
            165.63
                        CA
[]: ak.head(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
        ins_losses abbrev
     0
            145.08
                        AL
     1
            133.93
                        AK
[]: ak.tail(8)
[]:
         total
                speeding
                           alcohol not_distracted no_previous
                                                                   ins_premium \
     43
          19.4
                    7.760
                             7.372
                                             17.654
                                                                        1004.75
                                                           16.878
     44
          11.3
                    4.859
                             1.808
                                              9.944
                                                                         809.38
                                                           10.848
     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
     43
             156.83
                         TX
                         UT
     44
             109.48
     45
             109.61
                         VT
     46
             153.72
                         VA
     47
             111.62
                         WΑ
     48
             152.56
                         WV
```

```
49
             106.62
                         WI
     50
             122.04
                         WY
[]: ak.tail()
[]:
         total
                 speeding
                           alcohol
                                     not_distracted
                                                     no_previous
                                                                    ins_premium
     46
          12.7
                    2.413
                              3.429
                                              11.049
                                                            11.176
                                                                          768.95
     47
          10.6
                    4.452
                              3.498
                                               8.692
                                                                          890.03
                                                             9.116
     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
                                                                          791.14
                              5.568
                                              14.094
                                                            15.660
         ins_losses abbrev
     46
             153.72
                         VA
     47
             111.62
                         WA
     48
             152.56
                         WV
     49
             106.62
                         WΙ
     50
             122.04
                         WY
Г1:
     ak.shape
[]: (51, 8)
     ak.describe()
[]:
[]:
                 total
                         speeding
                                                                 no_previous
                                      alcohol
                                                not_distracted
            51.000000
                        51.000000
                                    51.000000
                                                                   51.000000
     count
                                                     51.000000
     mean
            15.790196
                         4.998196
                                     4.886784
                                                     13.573176
                                                                   14.004882
     std
                         2.017747
                                     1.729133
             4.122002
                                                      4.508977
                                                                    3.764672
     min
             5.900000
                         1.792000
                                     1.593000
                                                      1.760000
                                                                    5.900000
     25%
            12.750000
                         3.766500
                                     3.894000
                                                     10.478000
                                                                    11.348000
     50%
            15.600000
                         4.608000
                                     4.554000
                                                     13.857000
                                                                   13.775000
                                     5.604000
     75%
            18.500000
                         6.439000
                                                     16.140000
                                                                   16.755000
            23.900000
                         9.450000
                                    10.038000
                                                     23.661000
                                                                   21.280000
     max
            ins_premium
                          ins_losses
              51.000000
                           51.000000
     count
     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
[]: corr = ak.corr()
     corr
```

<ipython-input-12-cd014e0cc39d>: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.

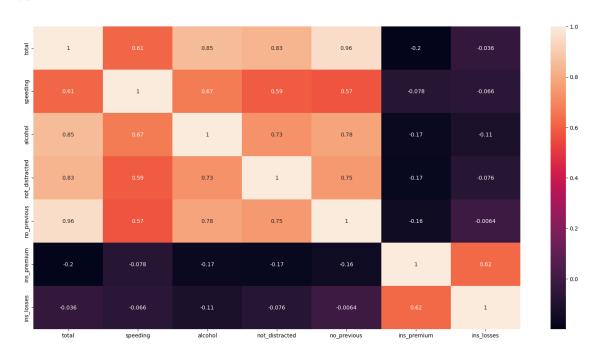
corr = ak.corr()

[]:		total	speeding	alcohol	${\tt not\_distracted}$	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_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

```
[]: plt.subplots(figsize = (20,10))
sns.heatmap(corr,annot=True)
```

## []: <Axes: >



#### []: ak["total"].value\_counts() []: 14.1 2 12.8 2 13.6 2 21.4 2 19.4 2 23.9 2 14.9 1 14.7 1 11.6 1 11.2 1 18.4 1 12.3 1 16.8 1 19.9 1 17.6 1 18.2 1 11.1 1 19.5 1 11.3 1 12.7 1 10.6 1 23.8 1 13.8 1 16.1 1 18.8 1 9.6 1 18.1 1 18.6 1 22.4 1 12.0 1 10.8 1 16.2 1 5.9 1 17.9 1 15.6 1 17.5 1 15.3 1 14.5 1 15.7 1 17.8 1 20.5 1

15.1

12.5

1

1

## 8.2 1 17.4 1

Name: total, dtype: int64

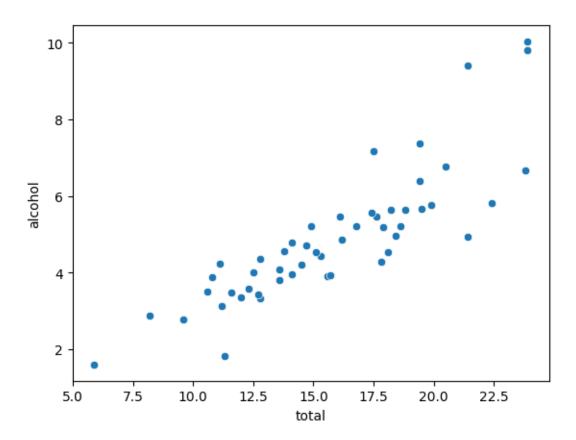
# []: ak.alcohol.value\_counts()

```
[]: 5.208
                2
     5.640
                1
     4.218
                1
     4.704
                1
     3.480
                1
     3.136
                1
     4.968
                1
     3.567
                1
     10.038
                1
     4.794
                1
     5.771
                1
     3.328
                1
     5.642
                1
     9.799
                1
     9.416
                1
     6.402
                1
     5.655
                1
     7.372
                1
     1.808
                1
     4.080
                1
     3.429
                1
     3.498
                1
     6.664
                1
     4.554
                1
     5.215
                1
     5.474
                1
     4.525
                1
     5.456
                1
     5.824
                1
     3.360
                1
     3.808
                1
     3.888
                1
     4.860
                1
     1.593
                1
     5.191
                1
     3.900
                1
     7.175
                1
     4.437
                1
     4.352
                1
     4.205
                1
```

3.925

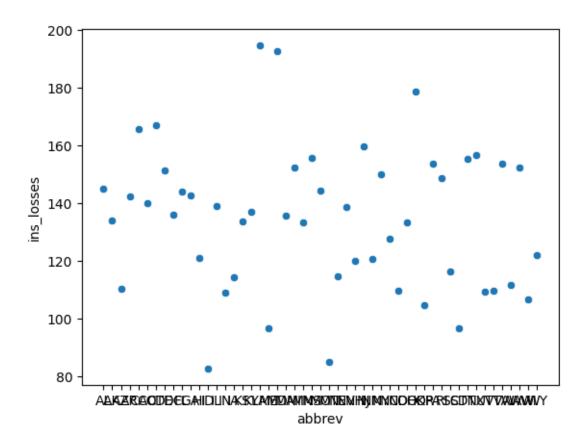
1

```
4.272
               1
     4.922
               1
     6.765
               1
     4.530
               1
     4.000
               1
     2.870
               1
     3.948
               1
     2.784
               1
               1
     5.568
     Name: alcohol, dtype: int64
[]: ak.isnull().any()
[]: total
                       False
     speeding
                       False
     alcohol
                       False
     not_distracted
                       False
     no_previous
                       False
     ins_premium
                       False
     ins_losses
                       False
     abbrev
                       False
     dtype: bool
[]: ak.isnull().sum()
[]: total
                       0
                       0
     speeding
     alcohol
                       0
     {\tt not\_distracted}
                       0
     no_previous
                       0
     ins_premium
                       0
     ins_losses
                       0
     abbrev
                       0
     dtype: int64
    DATA VISUALIZAION
[]: sns.scatterplot(x="total",y="alcohol", data=ak)
[]: <Axes: xlabel='total', ylabel='alcohol'>
```



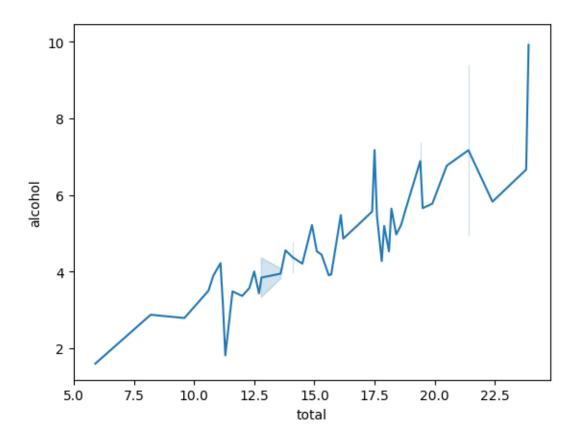
```
[]: sns.scatterplot(x="abbrev",y="ins_losses",data=ak)
```

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



```
[]: #Lineplot sns.lineplot(y="alcohol",x="total",data=ak)
```

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



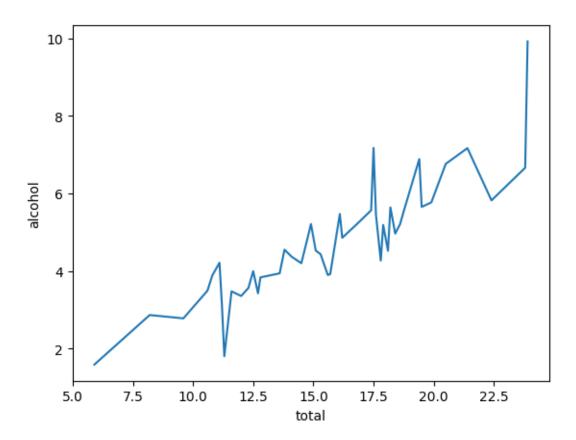
```
[]: #Lineplot sns.lineplot(y="alcohol",x="total",data=ak,ci=None)
```

<ipython-input-21-d491e3384d9c>:2: FutureWarning:

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

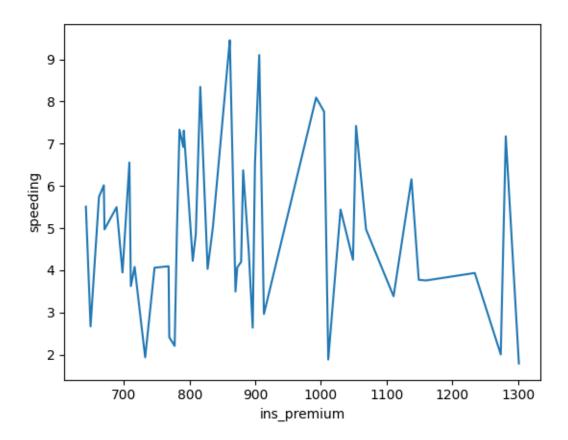
sns.lineplot(y="alcohol",x="total",data=ak,ci=None)

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



```
[]: sns.lineplot(x="ins_premium",y="speeding",data=ak)
```

[]: <Axes: xlabel='ins\_premium', ylabel='speeding'>



# []: #Displot sns.distplot(ak["total"])

<ipython-input-23-18a78abad740>:2: UserWarning:

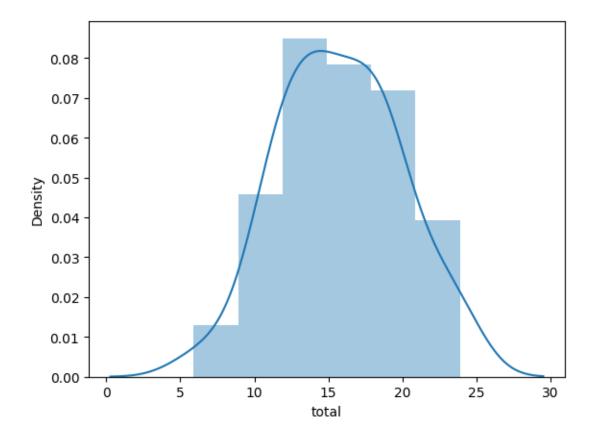
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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(ak["total"])

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



## []: sns.distplot(ak["not\_distracted"])

<ipython-input-24-bb0c8d7ed882>:1: UserWarning:

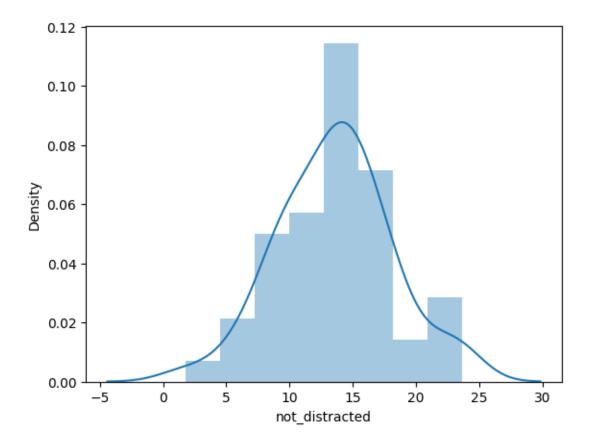
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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

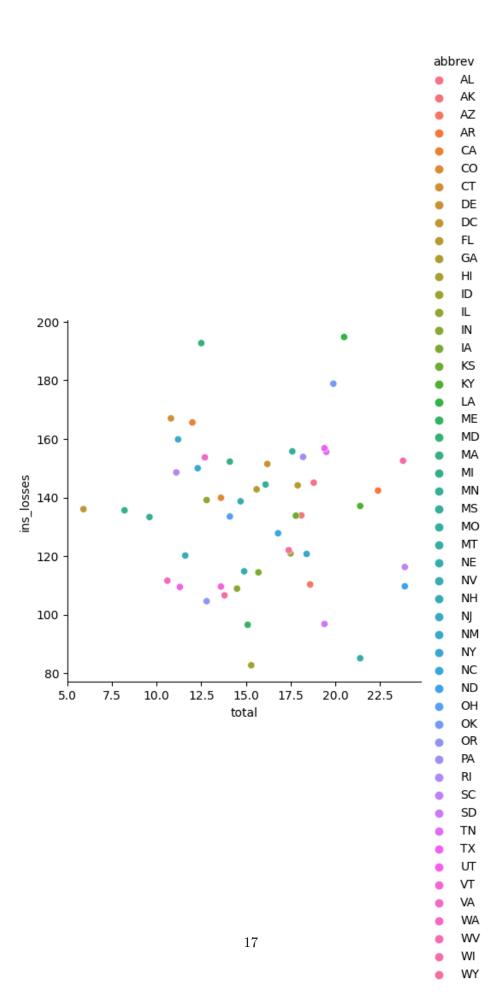
sns.distplot(ak["not\_distracted"])

[]: <Axes: xlabel='not\_distracted', ylabel='Density'>



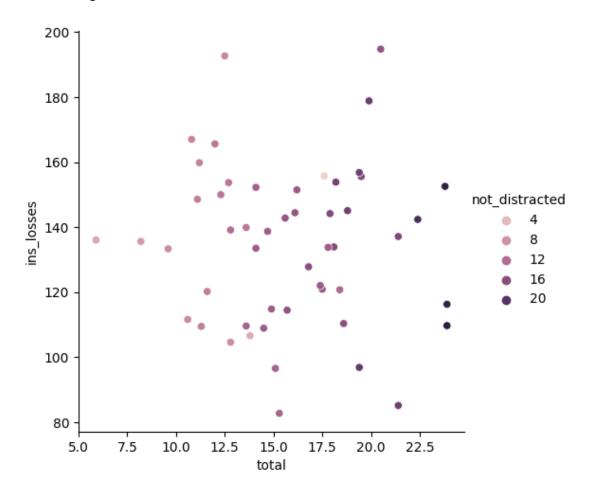
```
[]: #Relationplot sns.relplot(x="total",y="ins_losses",data=ak,hue="abbrev")
```

[]: <seaborn.axisgrid.FacetGrid at 0x7e3d2e3066b0>



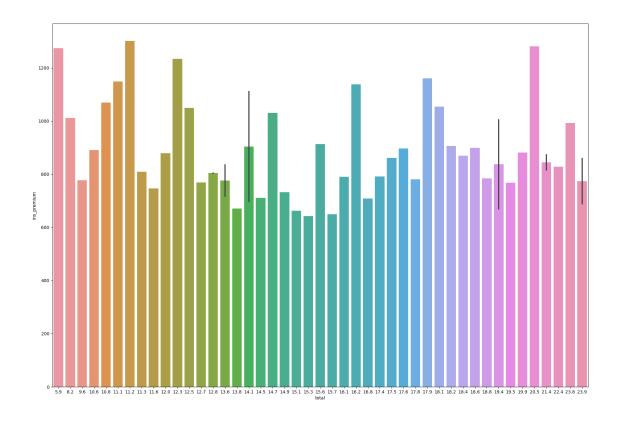
```
[]: #Relationplot
sns.relplot(x="total",y="ins_losses",data=ak,hue="not_distracted")
```

[]: <seaborn.axisgrid.FacetGrid at 0x7e3d2e496fe0>



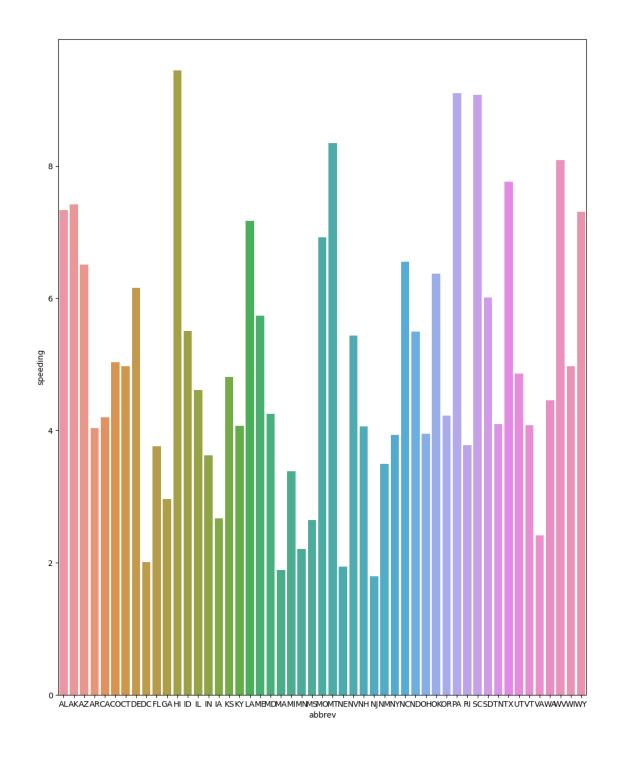
```
[]: #barplot
plt.subplots(figsize=(22,15))
sns.barplot(x="total",y="ins_premium",data=ak)
```

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



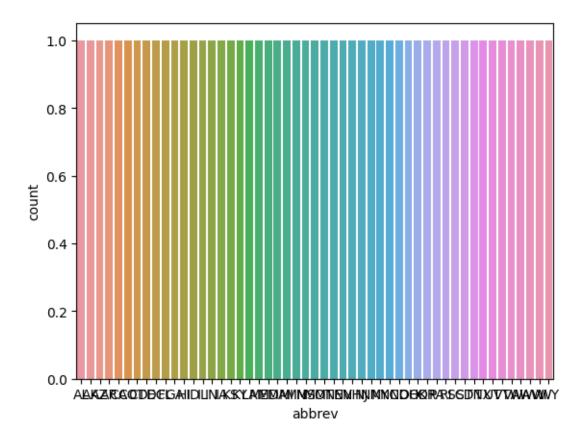
```
[]: #barplot
plt.subplots(figsize=(12,15))
sns.barplot(y="speeding",x="abbrev",data=ak)
```

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



```
[]: #Count Plot
sns.countplot(x="abbrev",data=ak)
```

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



```
[]: #boxplot
plt.subplots(figsize=(22,15))
sns.boxplot(x="total",y="ins_losses",data=ak)
```

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

