ASSIGNMENT-2

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```
pip install seaborn #library installed
Requirement already satisfied: seaborn in c:\users\tejos\anaconda\lib\
site-packages (0.12.2)
Requirement already satisfied: numpy!=1.24.0,>=1.17 in c:\users\tejos\
anaconda\lib\site-packages (from seaborn) (1.24.3)
Reguirement already satisfied: pandas>=0.25 in c:\users\tejos\
anaconda\lib\site-packages (from seaborn) (1.5.3)
Requirement already satisfied: matplotlib!=3.6.1,>=3.1 in c:\users\
tejos\anaconda\lib\site-packages (from seaborn) (3.7.1)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\tejos\
anaconda\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn)
(1.0.5)
Requirement already satisfied: cycler>=0.10 in c:\users\tejos\
anaconda\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn)
(0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\tejos\
anaconda\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn)
Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\tejos\
anaconda\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn)
(1.4.4)
Requirement already satisfied: packaging>=20.0 in c:\users\tejos\
anaconda\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn)
(23.0)
Requirement already satisfied: pillow>=6.2.0 in c:\users\tejos\
anaconda\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn)
(9.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\tejos\
anaconda\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn)
(3.0.9)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\tejos\
anaconda\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn)
(2.8.2)
Requirement already satisfied: pytz>=2020.1 in c:\users\tejos\
anaconda\lib\site-packages (from pandas>=0.25->seaborn) (2022.7)
Requirement already satisfied: six>=1.5 in c:\users\tejos\anaconda\
lib\site-packages (from python-dateutil>=2.7->matplotlib!=3.6.1,>=3.1-
>seaborn) (1.16.0)
Note: you may need to restart the kernel to use updated packages.
import matplotlib.pyplot as plt #import the matplotlib.pyplot module
and alias it as plt.
```

import seaborn as sns #import the seaborn library and alias it as
'sns'

print(sns.get_dataset_names()) #car_crashes is the inbuilt dataset
which we import

['anagrams', 'anscombe', 'attention', 'brain_networks', 'car_crashes',
'diamonds', 'dots', 'dowjones', 'exercise', 'flights', 'fmri',
'geyser', 'glue', 'healthexp', 'iris', 'mpg', 'penguins', 'planets',
'seaice', 'taxis', 'tips', 'titanic']

df=sns.load_dataset('car_crashes') # load the 'car_crashes' dataset
from seaborn and assign it to the variable 'df'

df #'df' contains the 'car_crashes' dataset, which provides
information about car crash statistics.

	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	13.6	5.032	3.808	10.744	12.920	835.50
6	10.8	4.968	3.888	9.396	8.856	1068.73
7	16.2	6.156	4.860	14.094	16.038	1137.87
8	5.9	2.006	1.593	5.900	5.900	1273.89
9	17.9	3.759	5.191	16.468	16.826	1160.13
10	15.6	2.964	3.900	14.820	14.508	913.15
11	17.5	9.450	7.175	14.350	15.225	861.18
12	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	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

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	6.923	5.474	14.812	13.524	790.32
26	21.4	8.346	9.416	17.976	18.190	816.21
27	14.9	1.937	5.215	13.857	13.410	732.28
28	14.7	5.439	4.704	13.965	14.553	1029.87
29	11.6	4.060	3.480	10.092	9.628	746.54
30	11.2	1.792	3.136	9.632	8.736	1301.52
31	18.4	3.496	4.968	12.328	18.032	869.85
32	12.3	3.936	3.567	10.824	9.840	1234.31
33	16.8	6.552	5.208	15.792	13.608	708.24
34	23.9	5.497	10.038	23.661	20.554	688.75
35	14.1	3.948	4.794	13.959	11.562	697.73
36	19.9	6.368	5.771	18.308	18.706	881.51
37	12.8	4.224	3.328	8.576	11.520	804.71
38	18.2	9.100	5.642	17.472	16.016	905.99
39	11.1	3.774	4.218	10.212	8.769	1148.99
40	23.9	9.082	9.799	22.944	19.359	858.97
41	19.4	6.014	6.402	19.012	16.684	669.31
42	19.5	4.095	5.655	15.990	15.795	767.91

43	19.4	7.760	7.372	17.654	16.878	1004.75
44	11.3	4.859	1.808	9.944	10.848	809.38
45	13.6	4.080	4.080	13.056	12.920	716.20
46	12.7	2.413	3.429	11.049	11.176	768.95
47	10.6	4.452	3.498	8.692	9.116	890.03
48	23.8	8.092	6.664	23.086	20.706	992.61
49	13.8	4.968	4.554	5.382	11.592	670.31
50	17.4	7.308	5.568	14.094	15.660	791.14
0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 29 20 20 20 20 20 20 20 20 20 20 20 20 20	ins_loss 145.1 133.1 110.1 142.1 165.1 136.1 151.1 136.1 144.1 120.8 82.1 139.1 108.1 114.1 133.1 137.1 194.1 96.1 152.1 135.1 155.1 144.8 85.1 114.1 85.1 138.1 159.1	93 AK 35 AZ 39 AR 63 CA 91 CO 02 CT 48 DE 05 DC 18 FL 80 GA 92 HI 75 ID 15 IL 92 IN 47 IA 80 KS 13 KY 78 LA 57 ME 70 MD 63 MA 26 MI 35 MN 77 MS 45 MO 15 NE 71 NV 21 NH				

```
31
        120.75
                    NM
32
        150.01
                    NY
33
        127.82
                    NC
34
        109.72
                    ND
35
        133.52
                    0H
36
        178.86
                    0K
37
        104.61
                    0R
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
                    VΑ
        153.72
47
        111.62
                    WA
48
        152.56
                    WV
49
                   WI
        106.62
50
        122.04
                   WY
df.info() # Display information about the DataFrame 'df', including
data types, non-null counts, and memory usage.
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 51 entries, 0 to 50
Data columns (total 8 columns):
                      Non-Null Count
#
     Column
                                      Dtype
- - -
 0
     total
                      51 non-null
                                       float64
1
     speeding
                      51 non-null
                                       float64
 2
                                       float64
     alcohol
                      51 non-null
 3
                      51 non-null
                                       float64
     not distracted
4
     no previous
                      51 non-null
                                       float64
 5
                                       float64
     ins premium
                      51 non-null
6
     ins losses
                      51 non-null
                                       float64
                      51 non-null
7
     abbrev
                                      object
dtypes: float64(7), object(1)
memory usage: 3.3+ KB
df.head(5) # Display the first 5 rows of the DataFrame 'df' to
provide an overview of its data.
   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
```

15.624

17.856

899.47

2

18.6

6.510

5.208

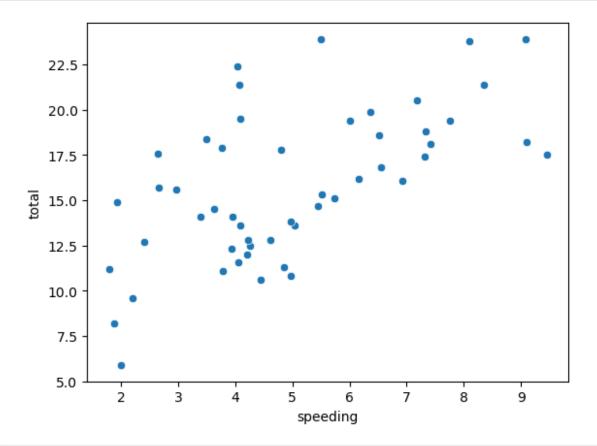
3	22.4	4.032	5.824	21.056	21.280	827.34
4	12.0	4.200	3.360	10.920	10.680	878.41

	ins_losses	abbrev
0	⁻ 145.08	AL
1	133.93	AK
2	110.35	AZ
3	142.39	AR
4	165.63	CA

sns.scatterplot(x="speeding",y="total",data=df) # Create a scatter
plot to visualize the relationship between speeding-related crashes
and total crashes.

As the number of speeding-related crashes increases, the total number of crashes also tends to increase

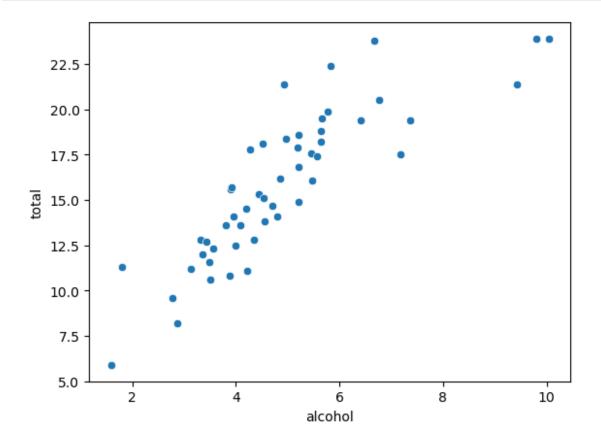
<Axes: xlabel='speeding', ylabel='total'>



sns.scatterplot(x="alcohol",y="total",data=df) # Create a scatter plot
to visualize the relationship between alcohol-related crashes and
total crashes.

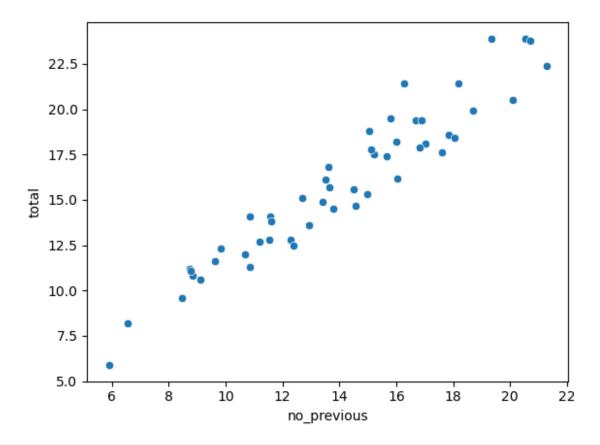
As the number of alcohol-related crashes increases, the total number of crashes also tends to increase

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

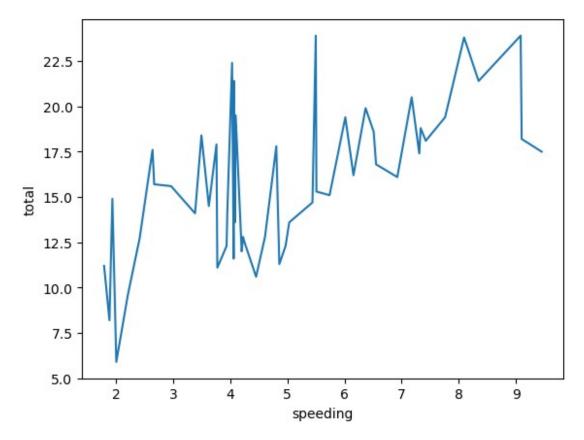


sns.scatterplot(x="no_previous",y="total",data=df)

<Axes: xlabel='no_previous', ylabel='total'>



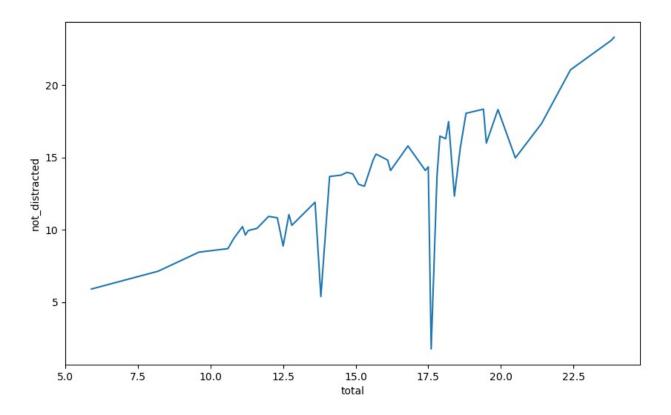
sns.lineplot(x="speeding",y="total",data=df,errorbar=None)
<Axes: xlabel='speeding', ylabel='total'>



```
#this line plots provides an initial visual insight into the
relationship between speeding and car crashes
#speed increases the total car crashes by the above plot

plt.figure(figsize=(10, 6))
sns.lineplot(data=df, x='total',
y='not_distracted',errorbar=None)#error bar is used since data
contains a string type data

<Axes: xlabel='total', ylabel='not_distracted'>
```



#The line plot suggests a relationship between the 'total' (possibly
total car crashes or incidents) and 'not_distracted' (possibly
incidents where there were no distracted accidents) variables
#as the increase in not_distracted increases the total car crashes
increases

sns.distplot(df["speeding"])

C:\Users\tejos\AppData\Local\Temp\ipykernel_18468\2127910581.py:1:
UserWarning:

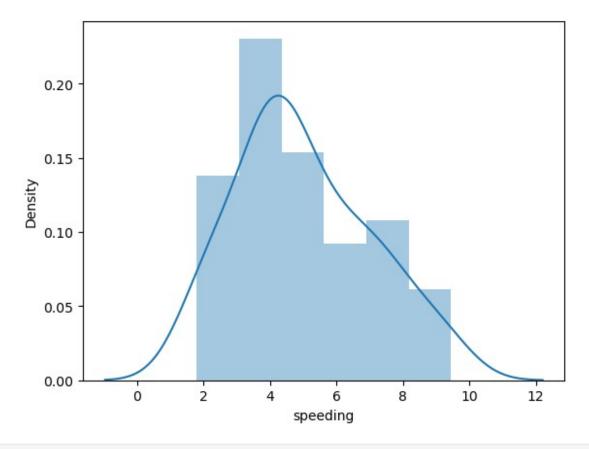
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

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

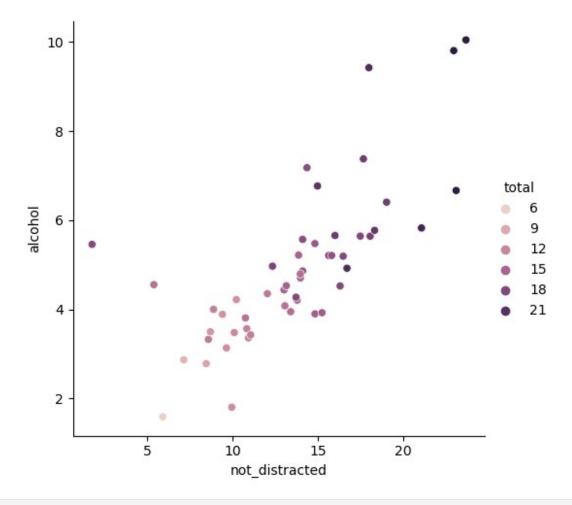
sns.distplot(df["speeding"])

<Axes: xlabel='speeding', ylabel='Density'>



#this plot helps you understand the distribution of the "speeding" data, including the central tendency(e.g., where the data tends to cluster) and any patterns, such as whether it follows a normal distribution, is skewed, or has multiple modes.
#data represented in the histogram is roughly symmetrically distributed around a central value, and the histogram takes on a shape that resembles a bell curve.
#this curve is bell shaped.

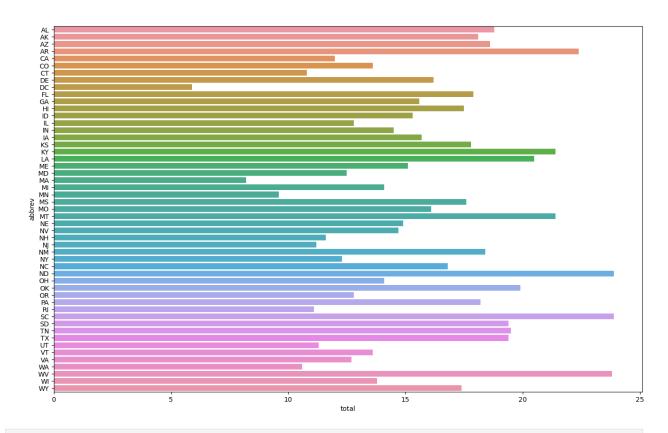
sns.relplot(x="not_distracted",y="alcohol",data=df,hue="total")
<seaborn.axisgrid.FacetGrid at 0x204df05f050>



#this plot will display the datapoints with "not_distracted" values on x-axis, "alcohol" values on y-axis, and use different colors to distinguish between different categories or values in the "total" column.

```
plt.subplots(figsize=(16,10))
sns.barplot(x='total',y= 'abbrev',data=df)
```

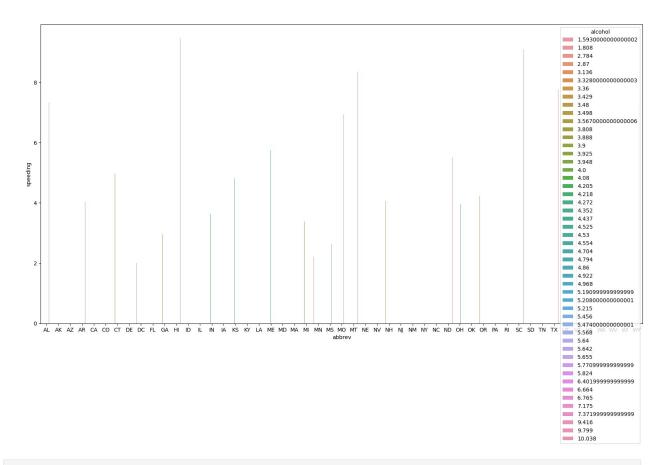
<Axes: xlabel='total', ylabel='abbrev'>



the resulting barplot will display bars for each unique value in the "abbrev" column(assuming 'abbrev'represents abbreviations or labels), with the height of each bar corresponding to the 'total'value associated with that abbreviation.

plt.figure(figsize=(20, 10)) #adjust the figure size as needed
sns.barplot(data=df, x='abbrev', y='speeding', hue='alcohol')

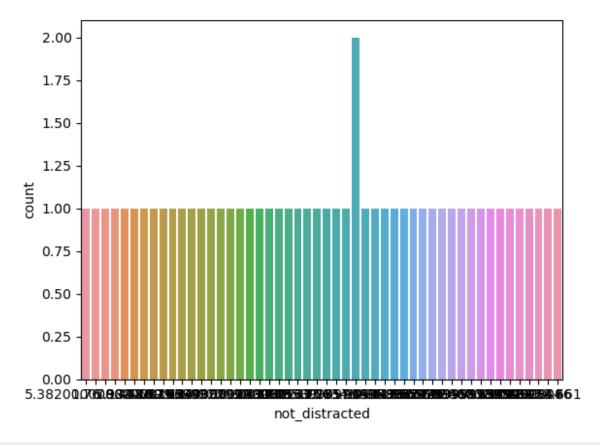
<Axes: xlabel='abbrev', ylabel='speeding'>



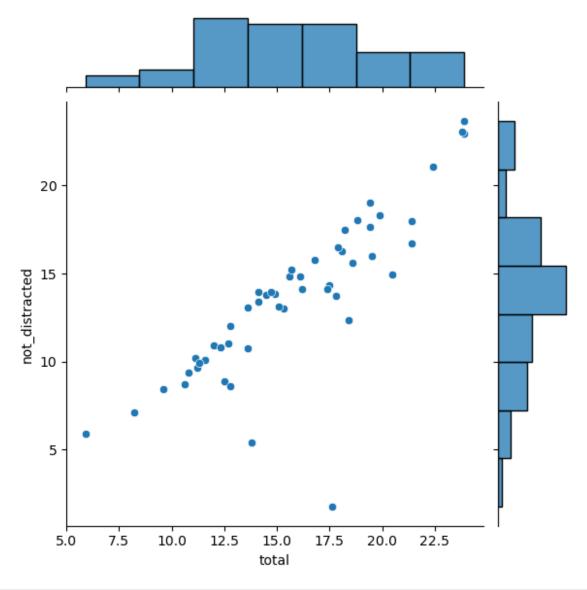
#creating a seaborn barplot with a larger figure size(width:20 units, height:10 units) to visualize the relationship between the 'speeding' column and the 'abbrev' column, with the 'alcohol' column for the hue.

sns.countplot(x="not_distracted",data=df)

<Axes: xlabel='not_distracted', ylabel='count'>



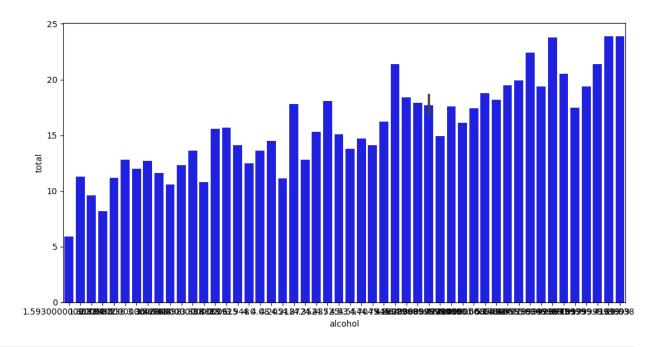
#the resulting countplot displays bars fo each unique value in the
"not_distracted" column, showing the frequency or count of each
category
#we can also see the same count in the value count.
sns.jointplot(x="total",y="not_distracted",data=df)
<seaborn.axisgrid.JointGrid at 0x204f20801d0>



```
#the resulting jointplot provides several insights into the
relationship between the "total" and "not_distracted" columns.

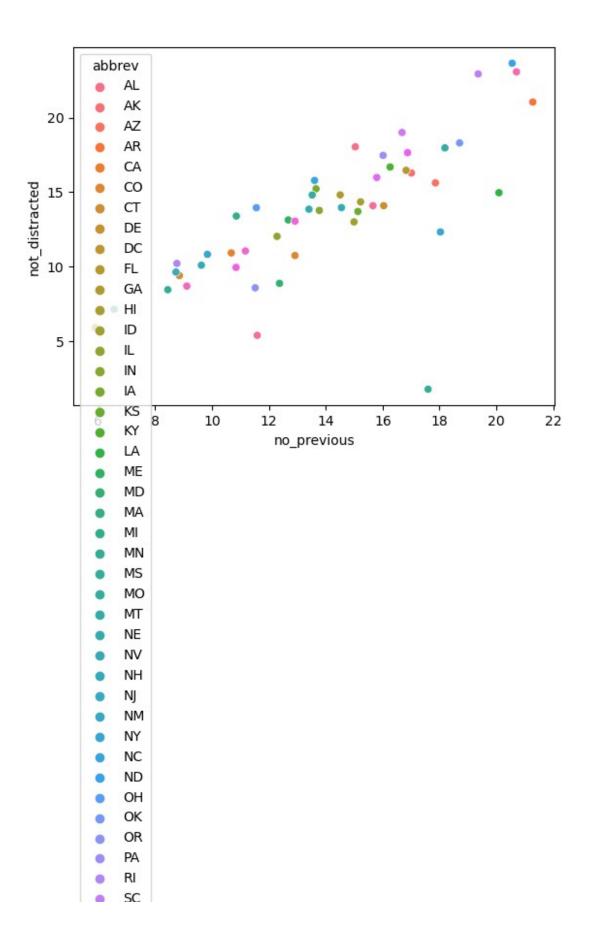
plt.figure(figsize=(12, 6))
sns.barplot(data=df, x='alcohol', y='total', color='blue',
label='Total Crashes')

<Axes: xlabel='alcohol', ylabel='total'>
```



#The resulting bar plot will display bars for each unique value in
the'alcohol' column, with the height of each bar representing the
'total'value associated with that abbreviation.
sns.scatterplot(data=df, x='no_previous', y='not_distracted',
hue='abbrev')
plt.xlabel('no_previous')

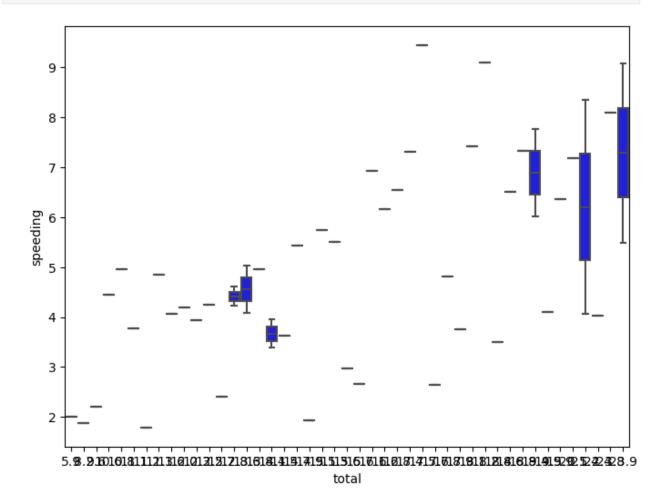
Text(0.5, 0, 'no_previous')



```
#The resulting scatter plot provides several insights into the
relationship between 'no_previous' and 'not_distracted,' with
consideration of the 'abbrev'

plt.figure(figsize=(8, 6))
sns.boxplot(data=df, x='total',y="speeding", color='blue')

<Axes: xlabel='total', ylabel='speeding'>
```

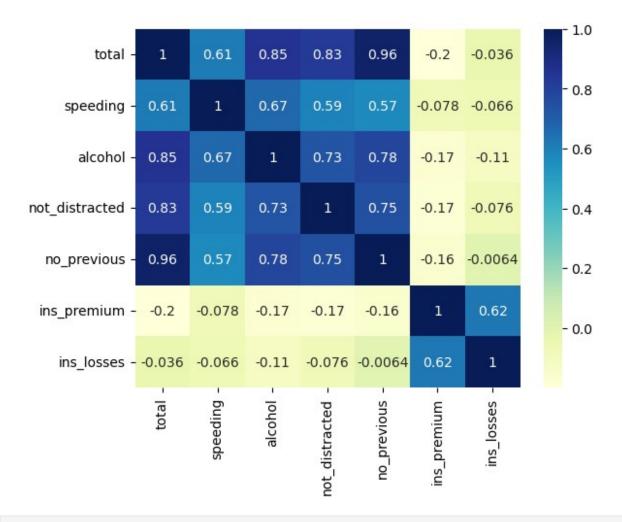


To analyze the relationship between the total number of crashes
('total') and the number of speed-related crashes ('speeding').

corr=df.corr()
corr

C:\Users\tejos\AppData\Local\Temp\ipykernel_18468\4178516133.py: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
            \
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
0.0\overline{0}6359
                ins premium
                             ins losses
total
                  -0.199702
                              -0.036011
                  -0.077675
                               -0.065928
speeding
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
#we can see that "no previous" is highly correlated to "total".we can
also see all the correlations
sns.heatmap(corr,annot=True,cmap="YlGnBu")
<Axes: >
```



#corellation between every attribute is given here
#>0.5 is highly correlated
#<0.5 is less correlated
#infernce:no_previous is highly collerated to total car crashes
sns.pairplot(df) #all the scatter plots and barplots
<seaborn.axisgrid.PairGrid at 0x204f4f3fdd0>

