ASSIGNMENT - 02

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Data Visualztion Exercises

1. Take car crashes dataset from seaborn library

2.load the dataset

3.Perfrom Data Visualization

4.Inference is must for each and every graph

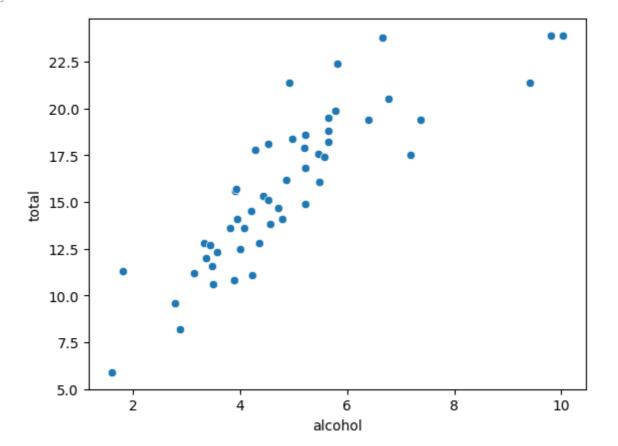
```
import numpy as np
In [ ]:
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
In [ ]: print(sns.get_dataset_names())
         ['anagrams', 'anscombe', 'attention', 'brain_networks', 'car_crashes', 'diamonds',
         'dots', 'dowjones', 'exercise', 'flights', 'fmri', 'geyser', 'glue', 'healthexp',
         'iris', 'mpg', 'penguins', 'planets', 'seaice', 'taxis', 'tips', 'titanic']
         df=sns.load_dataset('car_crashes')
         df.head()
Out[]:
            total speeding alcohol not_distracted no_previous ins_premium ins_losses abbrev
           18.8
                     7.332
                             5.640
                                          18.048
                                                       15.040
                                                                    784.55
                                                                              145.08
                                                                                         AL
                             4.525
                                           16.290
                                                       17.014
                                                                   1053.48
            18.1
                     7.421
                                                                              133.93
                                                                                         ΑK
         2 18.6
                     6.510
                             5.208
                                          15.624
                                                       17.856
                                                                    899.47
                                                                              110.35
                                                                                         ΑZ
            22.4
                     4.032
                             5.824
                                          21.056
                                                       21.280
                                                                    827.34
                                                                              142.39
                                                                                         AR
            12.0
                     4.200
                                                       10.680
                                                                    878.41
                                                                                         CA
                             3.360
                                          10.920
                                                                              165.63
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 51 entries, 0 to 50 Data columns (total 8 columns): Column Non-Null Count # Dtype ------------total 0 51 non-null float64 51 non-null float64 1 speeding alcohol 51 non-null float64 2 3 not_distracted 51 non-null float64 float64 4 no_previous 51 non-null 5 ins_premium 51 non-null float64 51 non-null float64 ins_losses 6 7 abbrev 51 non-null object dtypes: float64(7), object(1) memory usage: 3.3+ KB In []: df.shape (51, 8)Out[]: df.isnull().sum() In []: 0 total Out[]: speeding 0 0 alcohol not distracted 0 0 no_previous 0 ins_premium ins losses 0 abbrev 0 dtype: int64 df.describe() In []: Out[]: speeding alcohol not_distracted no_previous ins_premium ins_losses total **count** 51.000000 51.000000 51.000000 51.000000 51.000000 51.000000 51.000000 mean 15.790196 4.998196 4.886784 13.573176 14.004882 886.957647 134.493137 4.122002 2.017747 1.729133 4.508977 3.764672 178.296285 24.835922 std min 5.900000 1.792000 1.593000 1.760000 5.900000 641.960000 82.750000 10.478000 25% 12.750000 3.766500 3.894000 11.348000 768.430000 114.645000 50% 15.600000 4.608000 4.554000 13.857000 13.775000 858.970000 136.050000 **75%** 18.500000 6.439000 5.604000 16.140000 16.755000 1007.945000 151.870000 max 23.900000 9.450000 10.038000 23.661000 21.280000 1301.520000 194.780000

#total -> Number of drivers involved in fatal collisions per billion miles #speeding -> Percentage Of Drivers Involved In Fatal Collisions Who Were Speeding #alcohol -> Percentage Of Drivers Involved In Fatal Collisions Who Were Alcohol-Imp #not distracted -> Percentage Of Drivers Involved In Fatal Collisions Who Were Not #no previous -> Percentage Of Drivers Involved In Fatal Collisions Who Had Not Been #ins_premium -> Car Insurance Premiums #ins losses -> Losses incurred by insurance companies for collisions per insured di

Scatterplot

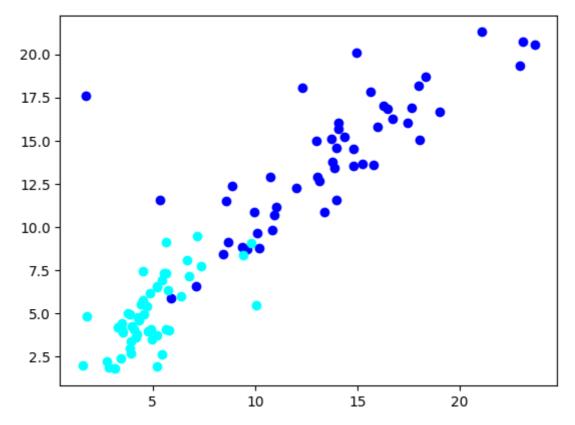
```
In [ ]: sns.scatterplot(x="alcohol",y="total",data=df)
Out[ ]: <Axes: xlabel='alcohol', ylabel='total'>
```



Inference: From the scatterplot it can be stated that when Percentage Of Drivers Involved In Fatal Collisions Who Were Alcohol-Impaired increases, Number of drivers involved in fatal collisions per billion miles also increases.

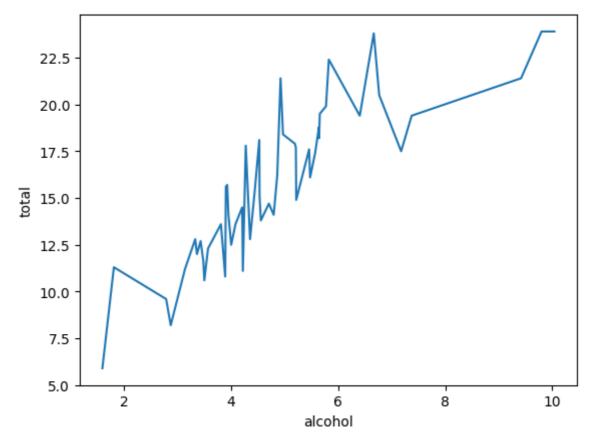
Comparing two plots

```
In [ ]: plt.scatter(x="not_distracted",y="no_previous",data=df,color="blue")
    plt.scatter(x="alcohol",y="speeding",data=df,color="cyan")
    plt.show()
```



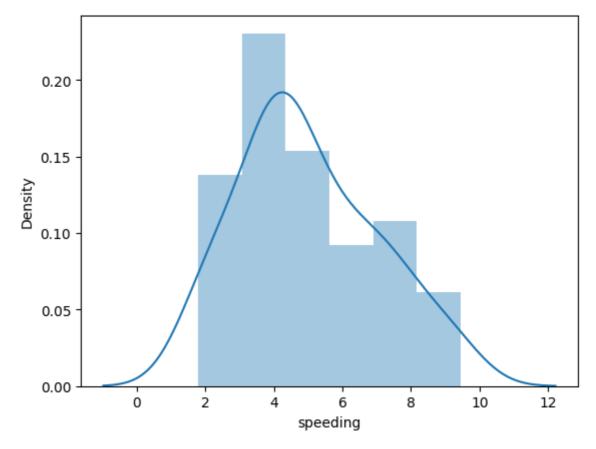
Inference: Comparing two scatterplots, one with accidents caused due to speeding and consumption of alcohol and two with accidents caused when driver was not distracted and drivers having no previous accident records.

Lineplot



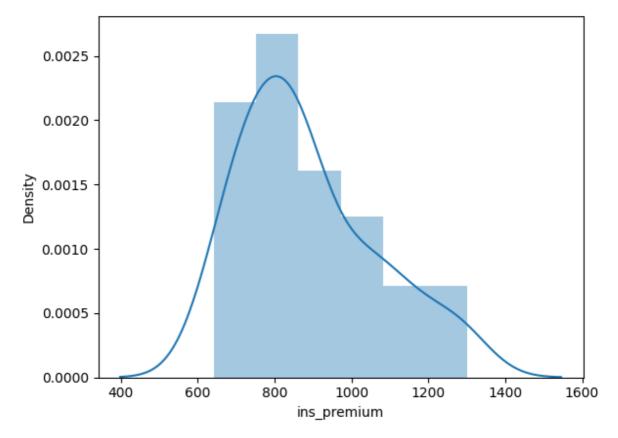
Inference: From the barplot it can be stated that as the number of alcohol impaired driver increases the total number od car crashes gradually increases.

Distribution plot

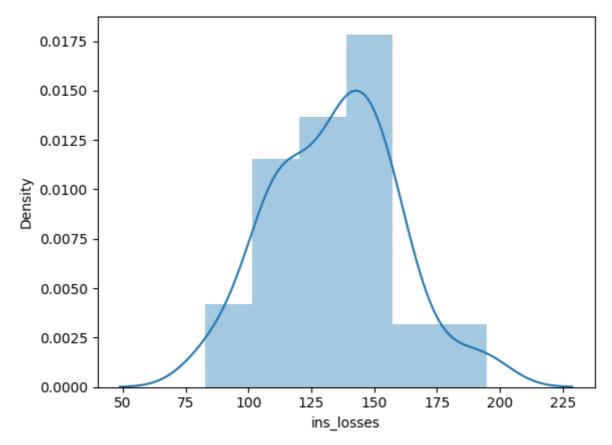


Inference: The above distribution plot shows the normal distribution of the variable "speeding" of the dataset.

file:///C:/Users/Rishima/Downloads/Assignment_02.html



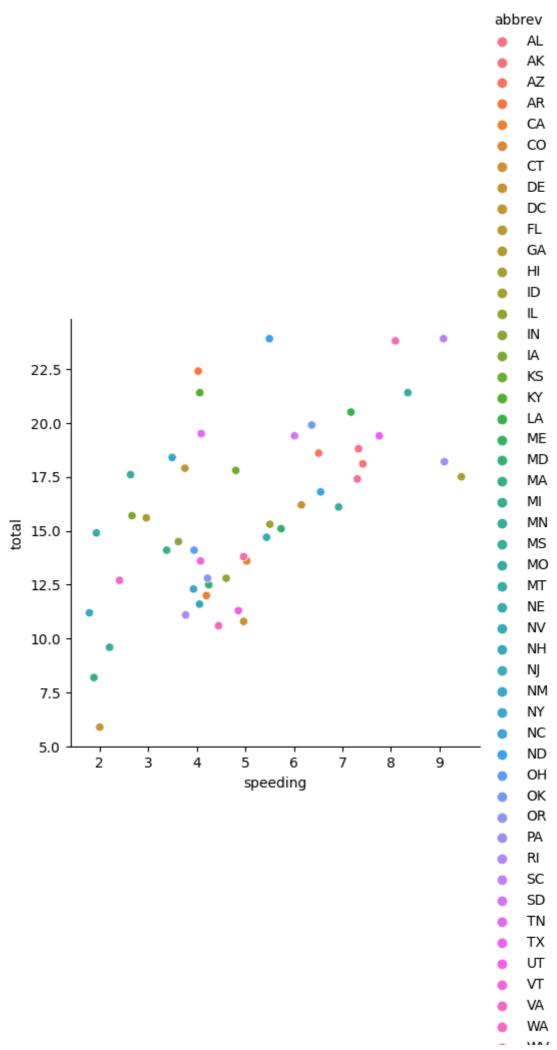
Inference: The above distribution plot shows the normal distribution of the variable "ins_premium" of the dataset.



Inference: The above distribution plot shows the normal distribution of the variable "ins_losses" of the dataset.

Relation plot

```
In [ ]: sns.relplot(x="speeding",y="total",data=df,hue="abbrev")
Out[ ]: <seaborn.axisgrid.FacetGrid at 0x79fe03303af0>
```

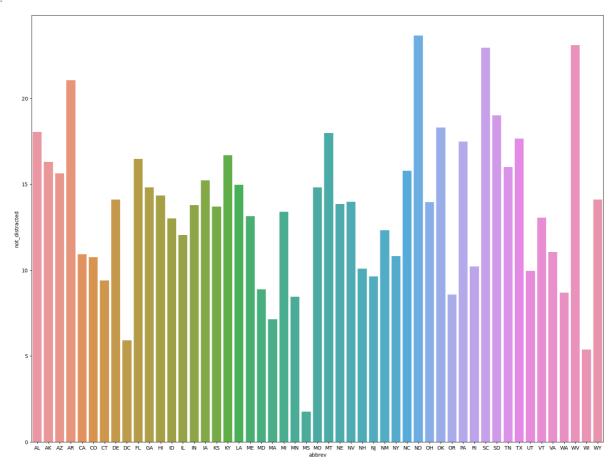


wvWIWY

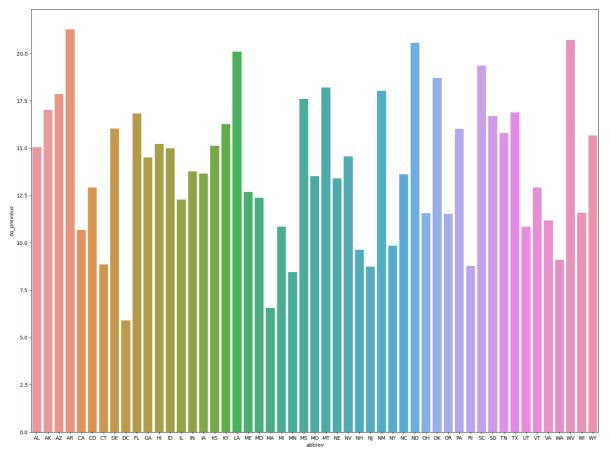
Inference: The relation plot depicts the proportional relation between crashes cause by high speeding and total crashes via different colour highlighting of the states.

```
df["abbrev"].value_counts()
Out[ ]:
                 1
          NV
                 1
         NH
                 1
                 1
         NJ
                 1
         NM
         NY
                 1
                 1
         NC
                 1
          ОН
                 1
         OK
                 1
         OR
                 1
          RΙ
                 1
         MT
                 1
          SC
                 1
          SD
                 1
                 1
          TN
          TX
                 1
         UT
                 1
                 1
          VT
                 1
         WA
                 1
         WV
                 1
         WI
                 1
         NE
                 1
         MO
                 1
          ΑK
          ID
                 1
          ΑZ
                 1
          AR
                 1
          CA
                 1
          CO
                 1
          \mathsf{CT}
                 1
         DE
                 1
         DC
                 1
          FL
                 1
         GΑ
                 1
         ΗI
                 1
          ΙL
                 1
         MS
                 1
          IN
                 1
          IΑ
                 1
          KS
                 1
          ΚY
                 1
          LA
                 1
         ME
                 1
         MD
                 1
                 1
         MA
          ΜI
                 1
                 1
         WY
          Name: abbrev, dtype: int64
```

Barplot



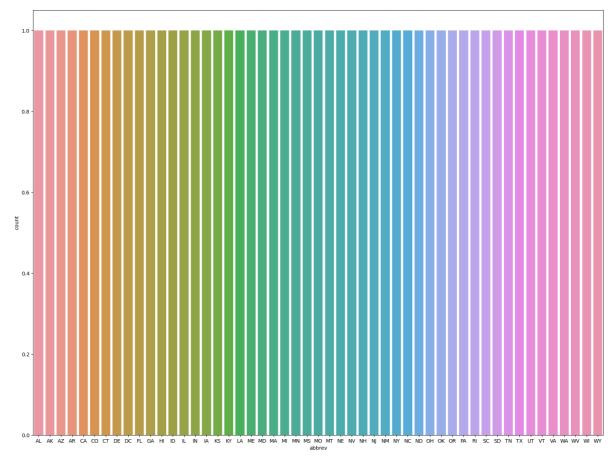
Inference: From the barplot it can be stated that the state abbreviated as ND has maximum Percentage Of Drivers Involved In Fatal Collisions Who Were Not Distracted and the state abbreviated MS has minimum Percentage Of Drivers Involved In Fatal Collisions Who Were Not Distracted.



Inference: From the barplot it can be stated that the state abbreviated as AR has maximum Percentage Of Drivers Involved In Fatal Collisions Who Had Not Been Involved In Any Previous Accidents and the state abbreviated DC has minimum Percentage Of Drivers Involved In Fatal Collisions Who Had Not Been Involved In Any Previous Accidents.

Countplot

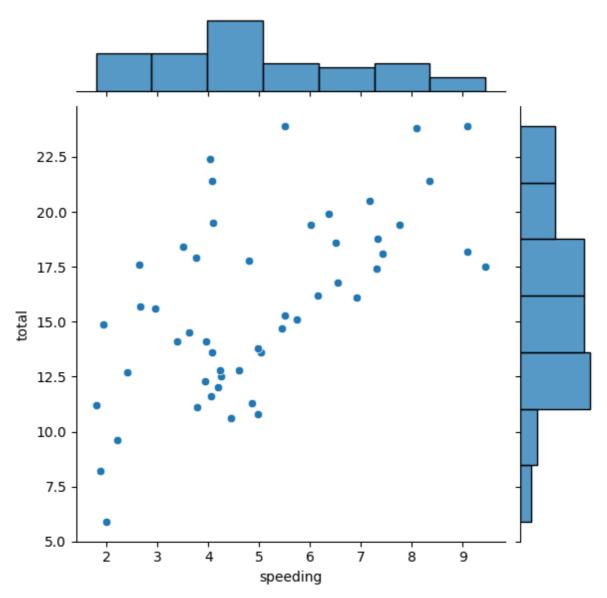
```
In [ ]: plt.subplots(figsize=(20,15))
    sns.countplot(x="abbrev",data=df)
Out[ ]: <Axes: xlabel='abbrev', ylabel='count'>
```



Inference: The above Countplot shows the count of occurence of the states' abbreviation in the dataset.

Jointplot

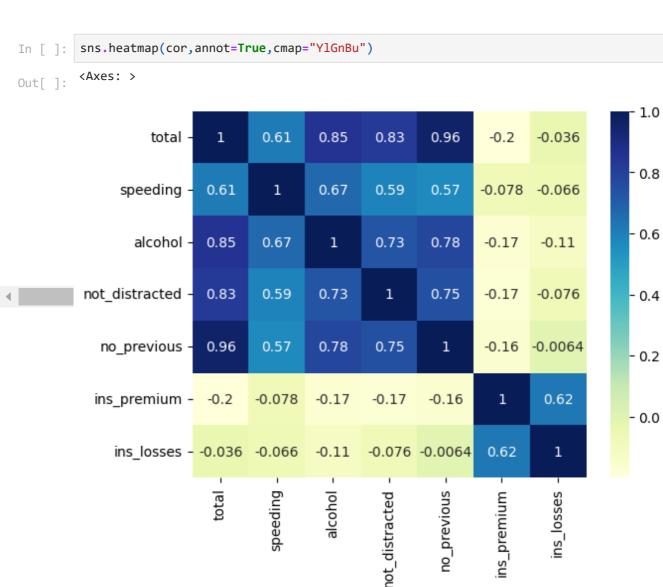
```
In [ ]: sns.jointplot(x="speeding",y="total",data=df)
Out[ ]: <seaborn.axisgrid.JointGrid at 0x79fe00dd3c40>
```



Inference: The relation between car crashes caused by speeding and total number of car crashes is depicted together in the jointplot.

Correlation and heatmap

Out[]:		total	speeding	alcohol	not_distracted	no_previous	ins_premium	ins_losse
	total	1.000000	0.611548	0.852613	0.827560	0.956179	-0.199702	-0.03601
	speeding	0.611548	1.000000	0.669719	0.588010	0.571976	-0.077675	-0.06592
	alcohol	0.852613	0.669719	1.000000	0.732816	0.783520	-0.170612	-0.11254
	not_distracted	0.827560	0.588010	0.732816	1.000000	0.747307	-0.174856	-0.07597
	no_previous	0.956179	0.571976	0.783520	0.747307	1.000000	-0.156895	-0.00635
	ins_premium	-0.199702	-0.077675	-0.170612	-0.174856	-0.156895	1.000000	0.62311
	ins_losses	-0.036011	-0.065928	-0.112547	-0.075970	-0.006359	0.623116	1.00000



Inference: The heatmap shows dependencies between two variables via correlation coefficients.