1. Take car crashes dataset from seaborn library

```
Requirement already satisfied: seaborn in /usr/local/lib/python3.10/dist-packages (0.12.2)
Requirement already satisfied: numpy!=1.24.0,>=1.17 in /usr/local/lib/python3.10/dist-packages (from seaborn) (1.23.5)
Requirement already satisfied: pandas>=0.25 in /usr/local/lib/python3.10/dist-packages (from seaborn) (1.5.3)
Requirement already satisfied: matplotlib!=3.6.1,>=3.1 in /usr/local/lib/python3.10/dist-packages (from seaborn) (3.7.1)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (1.1.0)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (4.42.1)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (1.4.5)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (23.1)
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (9.4.0)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (2.8.2)
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Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from pandas>=0.25->seaborn) (2023.3.post1)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from pandas>=0.25->seaborn) (2023.3.post1)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from pandas>=0.25->seaborn) (2023.3.post1)
```

2. load the dataset

```
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import rcParams
import seaborn as sns

df = sns.load_dataset('car_crashes')
df
```

```
total speeding alcohol not_distracted no_previous ins_premium ins_losses abbrev
      0
           18.8
                     7.332
                              5.640
                                              18.048
                                                            15.040
                                                                         784.55
                                                                                      145.08
                                                                                                       th
      1
           18.1
                    7.421
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           22.4
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                                              21.056
                                                           21.280
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                                              10.920
                                                            10.680
                                                                         878.41
                                                                                      165.63
                                                                                                 CA
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           13.6
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                              3.808
                                              10.744
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                                                            40 000
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df.info
     46
          12.7
                    2.413
                             3.429
                                             11.049
                                                           11.176
                                                                         768.95
                             3.498
                                                            9.116
                                                                         890.03
          10.6
                    4.452
                                              8.692
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                                                                         992.61
     48
          23.8
                    8.092
                             6.664
                                             23.086
                             4.554
                                              5.382
                                                           11.592
     49
          13.8
                    4.968
                                                                         670.31
     50
          17.4
                    7.308
                             5.568
                                             14.094
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         ins_losses abbrev
             145.08
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             133.80
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             137.13
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                         LA
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              96.57
                         ME
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             192.70
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             152.26
                         ΜI
     23
             133.35
     24
             155.77
                         MS
     25
             144.45
                         MO
     26
              85.15
     27
             114.82
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             138.71
                         NV
     29
             120.21
                         NH
     30
             159.85
                         NJ
             120.75
     32
             150.01
     33
             127.82
     34
             109.72
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             133.52
                         ОН
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                         OR
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             153.86
                         PA
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                         RΙ
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             155.57
                         TN
     43
             156.83
                         TX
     44
             109.48
                         UT
     45
             109.61
                         VT
             153.72
     47
             111.62
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                         WV
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df.head(10)
                                                                                                      Ш
         total speeding alcohol not_distracted no_previous ins_premium ins_losses abbrev
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                    7.332
                             5.640
                                             18.048
                                                          15.040
                                                                        784.55
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          18.1
                    7.421
                             4.525
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                                                          17.014
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          18.6
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          12.0
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                                                                        878.41
                                                                                    165.63
                                                                                                CA
          13.6
                    5.032
                             3.808
                                             10.744
                                                          12.920
                                                                        835.50
                                                                                    139.91
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                                              9.396
                                                           8.856
                                                                       1068.73
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                             4.860
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                                                          16.038
                                                                                    151.48
                                                                                                DE
                    6.156
```

3.Perfrom Data Visualization

5.9 17.9

1. Scatter plot

8

9

sns.scatterplot(x="alcohol", y="speeding", data=df)

2.006

3.759

1.593

5.191

₽

5.900

16.468

5.900

16.826

1273.89

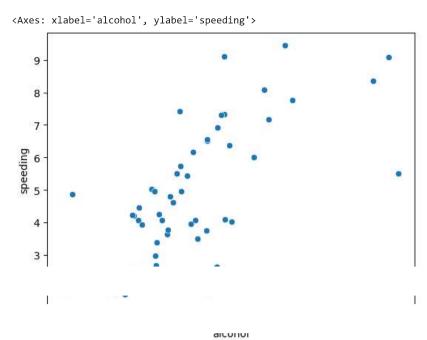
1160.13

136.05

144.18

DC

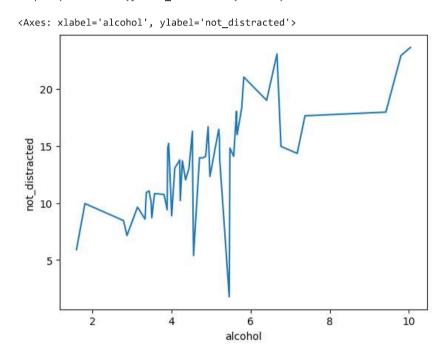
FL



Inference: In the above graph Speeding increases as alcohol increses. Alcohol is taken on x-axis and speeding is taken on y-axis.

2. Line Plot

sns.lineplot(x="alcohol",y="not_distracted",data=df)



Inference: Line plot with alcohol on x-axis and not_distracted on y-axis is plotted. In the graph not_distracted increases as alcohol increases.

3. Distribution Plot

0.02

no_previous

10

15

20

Inference: In this graph x-axis represents no_previous car crashes and y-axis represents the probability density. The density of previous car crashes increases until 15 and later decreases.

4. Relational plot

sns.relplot(x="total", y="ins_premium", data=df, hue="ins_losses")

| 1300 | 1200 | 1100 | 1200 | 1000 | 1200 | 1400 | 1600 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 | 1800 |

Inference: Relational plot with total on x-axis and ins_premium on y-axis with hue as ins_losses is plotted. It allows us to visualise how variables on a dataset relate to each other

5. Bar Plot

7.5

5.0

10.0

12.5

15.0 total 17.5

20.0

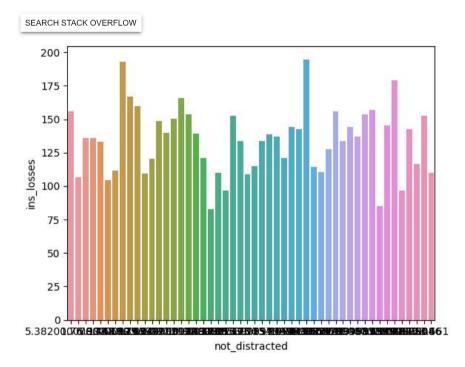
22.5

```
sns.barplot(data=df, x="not_distracted",y="ins_losses",errorbar=None)
sns.figure(figsize(17,8))
```

```
AttributeError Traceback (most recent call last)
<a href="mailto:right-26-56ed29c95608">right-26-56ed29c95608</a> in <cell line: 2>()

1 sns.barplot(data=df, x="not_distracted",y="ins_losses",errorbar=None)
---> 2 sns.figure(figsize(17,8))
```

AttributeError: module 'seaborn' has no attribute 'figure'

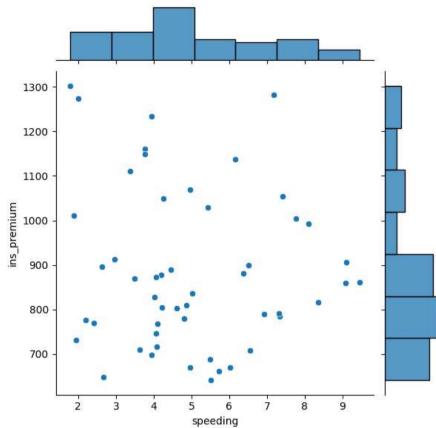


Inference: Bar graph is plotted with not_distracted on x-axis and ins_losses on y-axis.

6. Joint Plot

 $\verb|sns.jointplot(x="speeding", y="ins_premium", data=df)|\\$

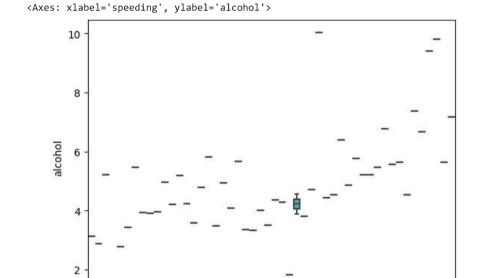
<seaborn.axisgrid.JointGrid at 0x7d1d20ebe320>



Inference: Joint plot with speeding on x axis and ins_premium on y-axis is plotted.

7. Box Plot

sns.boxplot(x="speeding",y="alcohol",data=df)



Inference: Box plot with speeding on x axis and alcohol on y-axis is plotted.

Correlation

cr=df.corr()
cr

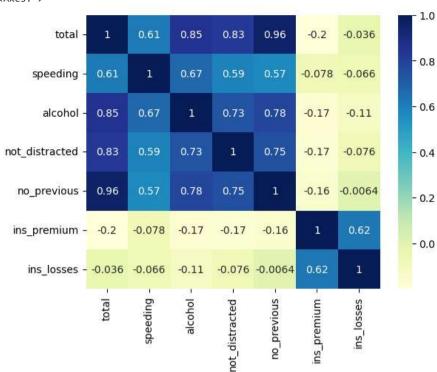
<ipython-input-19-5dc07b557697>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only
cr=df.corr()

	total	speeding	alcohol	not_distracted	no_previous	ins_premium	ins_losses	\blacksquare
total	1.000000	0.611548	0.852613	0.827560	0.956179	-0.199702	-0.036011	11.
speeding	0.611548	1.000000	0.669719	0.588010	0.571976	-0.077675	-0.065928	
alcohol	0.852613	0.669719	1.000000	0.732816	0.783520	-0.170612	-0.112547	
not_distracted	0.827560	0.588010	0.732816	1.000000	0.747307	-0.174856	-0.075970	
no_previous	0.956179	0.571976	0.783520	0.747307	1.000000	-0.156895	-0.006359	
ins_premium	-0.199702	-0.077675	-0.170612	-0.174856	-0.156895	1.000000	0.623116	
ins_losses	-0.036011	-0.065928	-0.112547	-0.075970	-0.006359	0.623116	1.000000	

8. Heat map

sns.heatmap(cr, annot=True,cmap="YlGnBu")

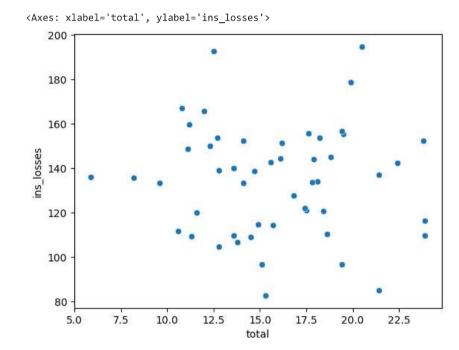
<Axes: >



Inference: Values above 0.5 are highly correlated and below 0.5 are less correlated.

9. Scatter plot

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



Inference: Scatter plot with total on x-axis and ins_losses on y-axis is plotted. In the graph ins_losses is maximum when total is 15.0 and 20.0.

10. Line plot

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

7.5

5.0

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



12

14

no_previous

16

18

20

22

10

✓ 0s completed at 11:00 PM

https://colab.research.google.com/drive/1j9TNTglr6r0-cLzVtb5w52GzvZTeEfjO#scrollTo=TjOP_MC-0aRZ&printMode=true

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