

# assignment2-21bce8601

September 13, 2023

```
[ ]: pip install seaborn
```

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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: pyparsing>=2.3.1 in  
/usr/local/lib/python3.10/dist-packages (from matplotlib!=3.6.1,>=3.1->seaborn)  
(3.1.1)  
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)  
Requirement already satisfied: pytz>=2020.1 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 python-dateutil>=2.7->matplotlib!=3.6.1,>=3.1->seaborn) (1.16.0)
```

```
[5]: import seaborn as sns

# Load the car crashes dataset
crashes = sns.load_dataset('car_crashes')
```

```
[ ]: crashes
```

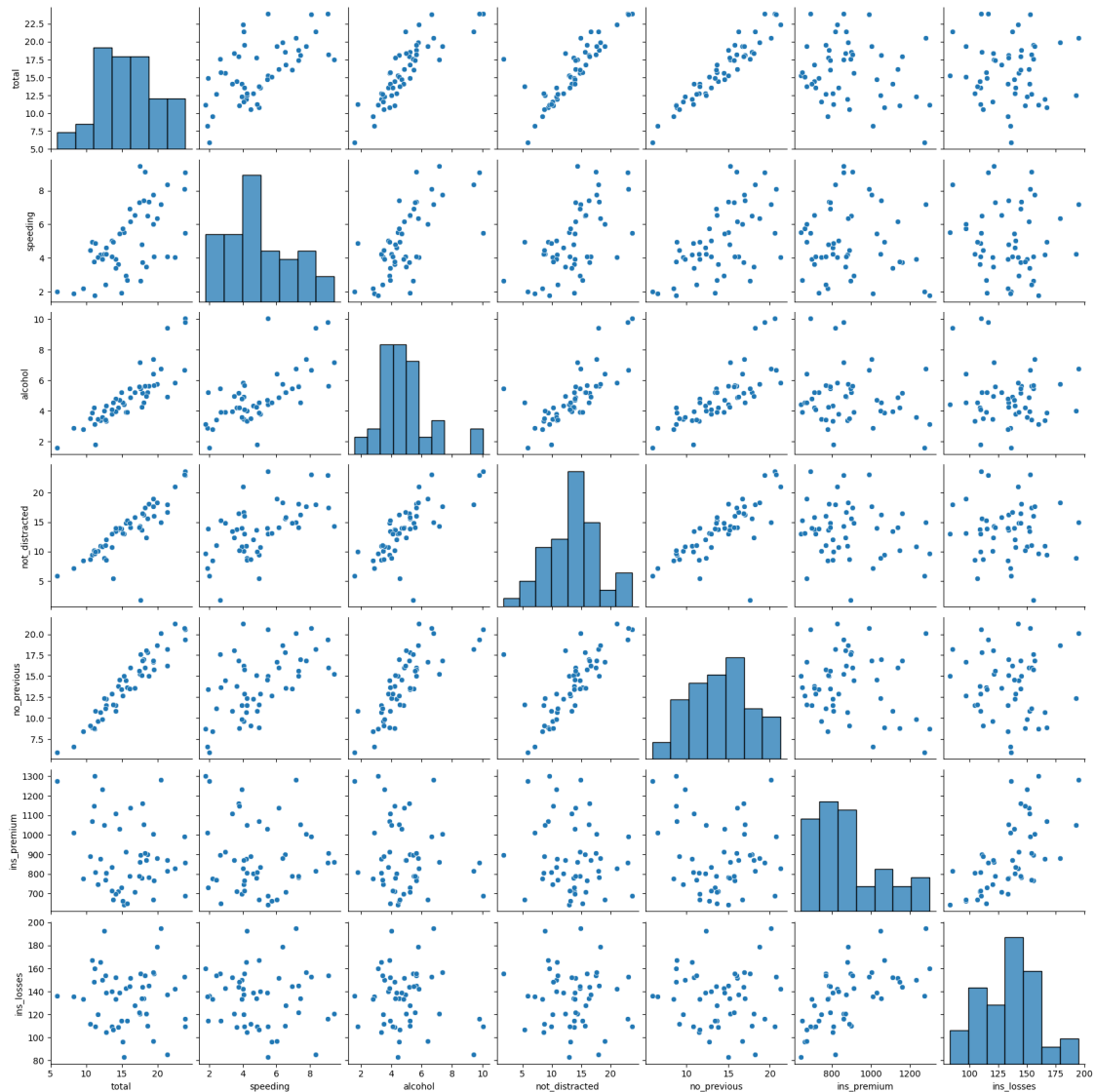
```
[ ]:      total  speeding  alcohol  not_distracted  no_previous  ins_premium  \
0      18.8      7.332    5.640          18.048          15.040          784.55
1      18.1      7.421    4.525          16.290          17.014         1053.48
2      18.6      6.510    5.208          15.624          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

	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
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

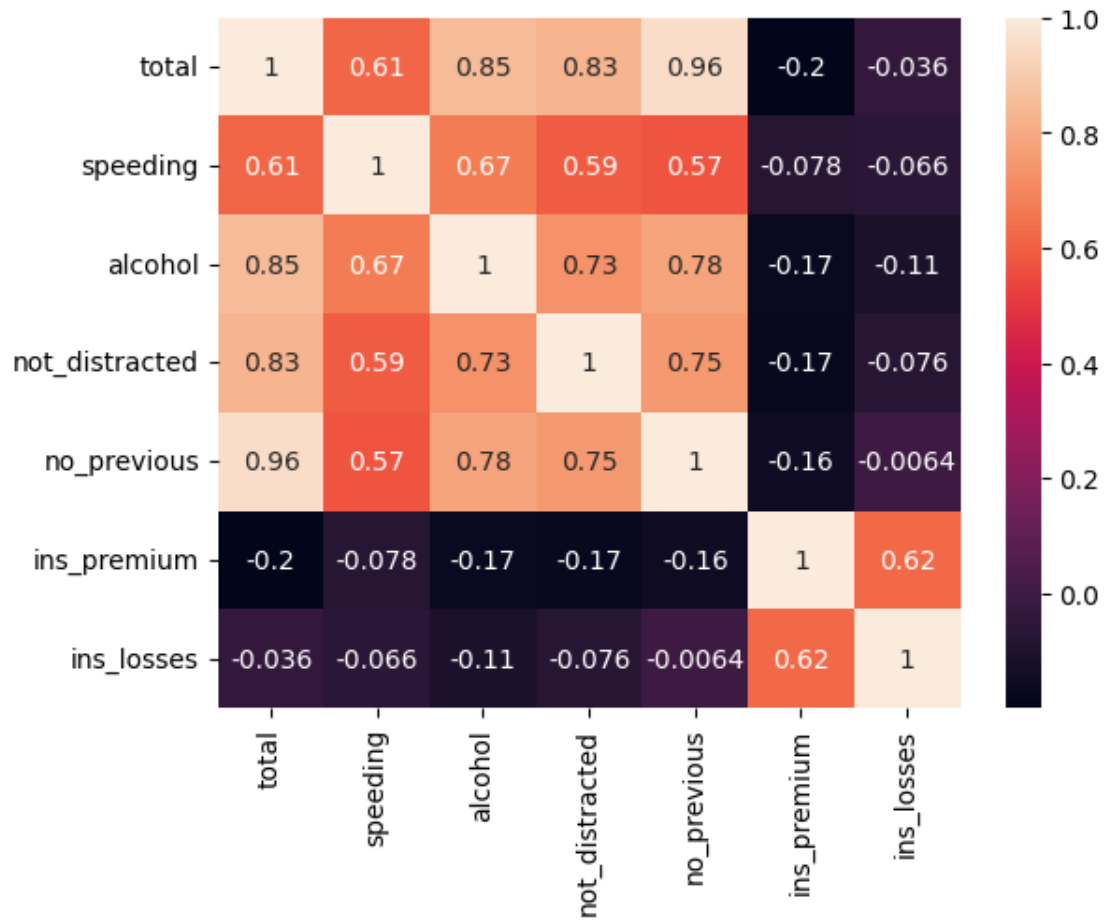
33	127.82	NC
34	109.72	ND
35	133.52	OH
36	178.86	OK
37	104.61	OR
38	153.86	PA
39	148.58	RI
40	116.29	SC
41	96.87	SD
42	155.57	TN
43	156.83	TX
44	109.48	UT
45	109.61	VT
46	153.72	VA
47	111.62	WA
48	152.56	WV
49	106.62	WI
50	122.04	WY

```
[6]: import matplotlib.pyplot as plt
sns.pairplot(crashes)
plt.show()
```



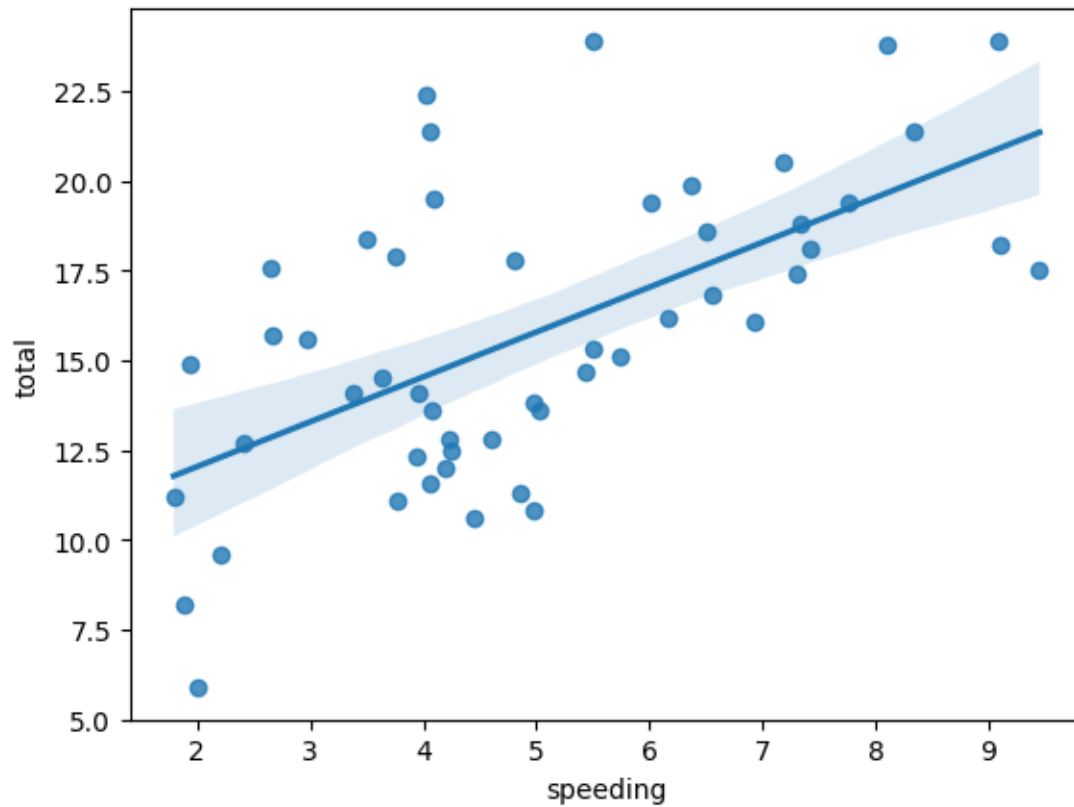
Inference: The pairplot allows us to visualize the relationships between all numeric variables in the dataset. We can see scatter plots for the numeric variables and histograms for the individual variables along the diagonal. It's useful for identifying potential correlations and distributions.

```
[8]: correlation = crashes.corr(numeric_only=True)
sns.heatmap(correlation, annot=True)
plt.show()
```



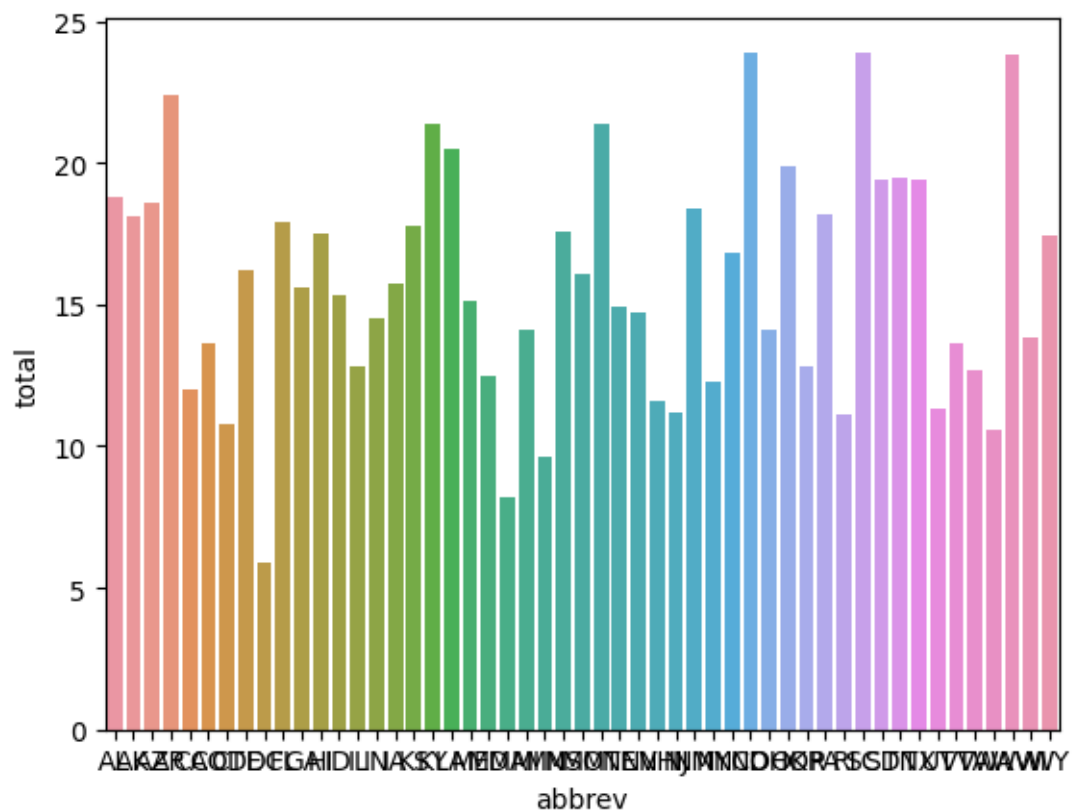
Inference: The heatmap shows the correlation between numeric variables. A value close to 1 indicates a strong positive correlation, while a value close to -1 indicates a strong negative correlation. This helps us understand which variables are most strongly related.

```
[9]: sns.regplot(x='speeding', y='total', data=crashes)
plt.show()
```



Inference: This regression plot shows the relationship between speeding and total number of crashes. It also includes a regression line which helps us understand the trend.

```
[10]: sns.barplot(x='abbrev', y='total', data=crashes)
plt.show()
```



Inference: This bar plot shows the total number of crashes for each state. It allows us to compare the crash counts between different states