```
In [ ]:
In [1]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
In [2]: sns.get_dataset_names()
Out[2]: ['anagrams',
          'anscombe',
          'attention',
          'brain_networks',
          'car_crashes',
          'diamonds',
          'dots',
          'dowjones',
          'exercise',
          'flights',
          'fmri',
          'geyser',
          'glue',
          'healthexp',
          'iris',
          'mpg',
          'penguins',
          'planets',
          'seaice',
          'taxis',
          'tips',
          'titanic']
```

In [4]: df=sns.load_dataset('car_crashes')
df

Out[4]:

:		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	AL
	1	18.1	7.421	4.525	16.290	17.014	1053.48	133.93	AK
	2	18.6	6.510	5.208	15.624	17.856	899.47	110.35	AZ
	3	22.4	4.032	5.824	21.056	21.280	827.34	142.39	AR
	4	12.0	4.200	3.360	10.920	10.680	878.41	165.63	CA
	5	13.6	5.032	3.808	10.744	12.920	835.50	139.91	CO
	6	10.8	4.968	3.888	9.396	8.856	1068.73	167.02	CT
	7	16.2	6.156	4.860	14.094	16.038	1137.87	151.48	DE
	8	5.9	2.006	1.593	5.900	5.900	1273.89	136.05	DC
	9	17.9	3.759	5.191	16.468	16.826	1160.13	144.18	FL
	10	15.6	2.964	3.900	14.820	14.508	913.15	142.80	GA
	11	17.5	9.450	7.175	14.350	15.225	861.18	120.92	HI
	12	15.3	5.508	4.437	13.005	14.994	641.96	82.75	ID
	13	12.8	4.608	4.352	12.032	12.288	803.11	139.15	IL
	14	14.5	3.625	4.205	13.775	13.775	710.46	108.92	IN
	15	15.7	2.669	3.925	15.229	13.659	649.06	114.47	IA
	16	17.8	4.806	4.272	13.706	15.130	780.45	133.80	KS
	17	21.4	4.066	4.922	16.692	16.264	872.51	137.13	KY
	18	20.5	7.175	6.765	14.965	20.090	1281.55	194.78	LA
	19	15.1	5.738	4.530	13.137	12.684	661.88	96.57	ME
	20	12.5	4.250	4.000	8.875	12.375	1048.78	192.70	MD
	21	8.2	1.886	2.870	7.134	6.560	1011.14	135.63	MA
	22	14.1	3.384	3.948	13.395	10.857	1110.61	152.26	MI
	23	9.6	2.208	2.784	8.448	8.448	777.18	133.35	MN
	24	17.6	2.640	5.456	1.760	17.600	896.07	155.77	MS
	25	16.1	6.923	5.474	14.812	13.524	790.32	144.45	МО
	26	21.4	8.346	9.416	17.976	18.190	816.21	85.15	MT
	27	14.9	1.937	5.215	13.857	13.410	732.28	114.82	NE
	28	14.7	5.439	4.704	13.965	14.553	1029.87	138.71	NV
	29	11.6	4.060	3.480	10.092	9.628	746.54	120.21	NH
	30	11.2	1.792	3.136	9.632	8.736	1301.52	159.85	NJ
	31	18.4	3.496	4.968	12.328	18.032	869.85	120.75	NM
	32	12.3	3.936	3.567	10.824	9.840	1234.31	150.01	NY

	total	speeding	alcohol	not_distracted	no_previous	ins_premium	ins_losses	abbrev
33	16.8	6.552	5.208	15.792	13.608	708.24	127.82	NC
34	23.9	5.497	10.038	23.661	20.554	688.75	109.72	ND
35	14.1	3.948	4.794	13.959	11.562	697.73	133.52	ОН
36	19.9	6.368	5.771	18.308	18.706	881.51	178.86	OK
37	12.8	4.224	3.328	8.576	11.520	804.71	104.61	OR
38	18.2	9.100	5.642	17.472	16.016	905.99	153.86	PA
39	11.1	3.774	4.218	10.212	8.769	1148.99	148.58	RI
40	23.9	9.082	9.799	22.944	19.359	858.97	116.29	SC
41	19.4	6.014	6.402	19.012	16.684	669.31	96.87	SD
42	19.5	4.095	5.655	15.990	15.795	767.91	155.57	TN
43	19.4	7.760	7.372	17.654	16.878	1004.75	156.83	TX
44	11.3	4.859	1.808	9.944	10.848	809.38	109.48	UT
45	13.6	4.080	4.080	13.056	12.920	716.20	109.61	VT
46	12.7	2.413	3.429	11.049	11.176	768.95	153.72	VA
47	10.6	4.452	3.498	8.692	9.116	890.03	111.62	WA
48	23.8	8.092	6.664	23.086	20.706	992.61	152.56	WV
49	13 ጸ	4 968	4 554	5 382	11 502	670 31	106 62	\\/I

In	[5]	:	df.	info
			J	

Out[5]:				me.info of	total	speeding	alcohol	not_distracted
		revious			10.0	40 4	F 040	704 55
	0	18.8	7.332	5.640	18.0		5.040	784.55
	1	18.1	7.421	4.525	16.2		7.014	1053.48
	2	18.6	6.510	5.208	15.6		7.856	899.47
	3	22.4	4.032	5.824	21.0		1.280	827.34
	4	12.0	4.200	3.360	10.9		0.680	878.41
	5	13.6	5.032	3.808	10.7		2.920	835.50
	6	10.8	4.968	3.888	9.3		8.856	1068.73
	7	16.2	6.156	4.860	14.0	94 1	6.038	1137.87
	8	5.9	2.006	1.593	5.9	00	5.900	1273.89
	9	17.9	3.759	5.191	16.4	68 1	6.826	1160.13
	10	15.6	2.964	3.900	14.8	20 1	4.508	913.15
	11	17.5	9.450	7.175	14.3	50 1	5.225	861.18
	12	15.3	5.508	4.437	13.0	0 5 1	4.994	641.96
	13	12.8	4.608	4.352	12.0	32 1	2.288	803.11
	14	14.5	3.625	4.205	13.7	75 1	3.775	710.46
	15	15.7	2.669	3.925	15.2	29 1	3.659	649.06
	16	17.8	4.806	4.272	13.7		5.130	780.45
	17	21.4	4.066	4.922	16.6		6.264	872.51
	18	20.5	7.175	6.765	14.9		0.090	1281.55
	19	15.1	5.738	4.530	13.1		2.684	661.88
	20	12.5	4.250	4.000	8.8		2.375	1048.78
	21	8.2	1.886	2.870	7.1		6.560	1011.14
	22	14.1	3.384	3.948	13.3		0.857	1110.61
	23	9.6	2.208	2.784	8.4		8.448	777.18
	24	17.6	2.640	5.456	1.7		7.600	896.07
	25	16.1	6.923	5.474	14.8		3.524	790.32
	26	21.4	8.346	9.416	17.9		8.190	816.21
	27	14.9	1.937	5.215	13.8		3.410	732.28
	28	14.7	5.439	4.704	13.9		4.553	1029.87
	29	11.6	4.060	3.480	10.0		9.628	746.54
	30	11.0	1.792	3.480				1301.52
		18.4			9.6 12.3		8.736	
	31		3.496 3.936	4.968	10.8		8.032 9.840	869.85
	32	12.3		3.567				1234.31
	33	16.8	6.552	5.208	15.7		3.608	708.24
	34	23.9	5.497	10.038	23.6		0.554	688.75
	35	14.1	3.948	4.794	13.9		1.562	697.73
	36	19.9	6.368	5.771	18.3		8.706	881.51
	37	12.8	4.224	3.328	8.5		1.520	804.71
	38	18.2	9.100	5.642	17.4		6.016	905.99
	39	11.1	3.774	4.218	10.2		8.769	1148.99
	40	23.9	9.082	9.799	22.9		9.359	858.97
	41	19.4	6.014	6.402	19.0		6.684	669.31
	42	19.5	4.095	5.655	15.9		5.795	767.91
	43	19.4	7.760	7.372	17.6		6.878	1004.75
	44	11.3	4.859	1.808	9.9		0.848	809.38
	45	13.6	4.080	4.080	13.0		2.920	716.20
	46	12.7	2.413	3.429	11.0	49 1	1.176	768.95
	47	10.6	4.452	3.498	8.6	92	9.116	890.03
	48	23.8	8.092	6.664	23.0	86 2	0.706	992.61
	49	13.8	4.968	4.554	5.3	82 1	1.592	670.31
	50	17.4	7.308	5.568	14.0	94 1	5.660	791.14

	ins_losses	ahhrev
0	145.08	AL
1	133.93	
2	110.35	AZ
3	142.39	AR
4	165.63	CA
5	139.91	
		CO CT
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 >

In [6]: df.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	<pre>not_distracted</pre>	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
44	C1+C4/7\	-1-2+/41	

dtypes: float64(7), object(1)

memory usage: 3.3+ KB

In [7]: df.describe()

Out[7]:

	total	speeding	alcohol	not_distracted	no_previous	ins_premium	ins_losses
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
std	4.122002	2.017747	1.729133	4.508977	3.764672	178.296285	24.835922
min	5.900000	1.792000	1.593000	1.760000	5.900000	641.960000	82.750000
25%	12.750000	3.766500	3.894000	10.478000	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

In [8]: df.head()

Out[8]:

	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	AL	
1	18.1	7.421	4.525	16.290	17.014	1053.48	133.93	AK	
2	18.6	6.510	5.208	15.624	17.856	899.47	110.35	AZ	
3	22.4	4.032	5.824	21.056	21.280	827.34	142.39	AR	
4	12.0	4.200	3.360	10.920	10.680	878.41	165.63	CA	

```
df.tail()
 In [9]:
 Out[9]:
                              alcohol not_distracted no_previous ins_premium ins_losses abbrev
               total speeding
           46
               12.7
                        2.413
                                3.429
                                             11.049
                                                         11.176
                                                                      768.95
                                                                                 153.72
                                                                                           VA
           47
               10.6
                        4.452
                                3.498
                                             8.692
                                                          9.116
                                                                      890.03
                                                                                 111.62
                                                                                           WA
               23.8
                        8.092
                                6.664
                                             23.086
                                                                                           WV
           48
                                                         20.706
                                                                      992.61
                                                                                 152.56
                                                                                           WI
           49
                        4.968
                                4.554
                                              5.382
                                                         11.592
                                                                      670.31
                                                                                 106.62
               13.8
                        7.308
                                                                                 122.04
                                                                                           WY
           50
               17.4
                                5.568
                                             14.094
                                                         15.660
                                                                      791.14
 In [ ]:
In [10]: df.isnull().any()
Out[10]: total
                               False
          speeding
                               False
          alcohol
                               False
          not_distracted
                               False
          no_previous
                               False
          ins_premium
                               False
          ins_losses
                               False
          abbrev
                               False
          dtype: bool
In [11]: df.isnull().sum()
Out[11]: total
                               0
                               0
          speeding
                               0
          alcohol
          not_distracted
                               0
                               0
          no_previous
                               0
          ins_premium
                               0
          ins_losses
                               0
          abbrev
          dtype: int64
In [12]:
          df.isna().sum()
Out[12]: total
                               0
                               0
          speeding
          alcohol
                               0
          not_distracted
                               0
                               0
          no_previous
                               0
          ins_premium
          ins_losses
                               0
          abbrev
                               0
          dtype: int64
```

In [13]: corr=df.corr()
corr

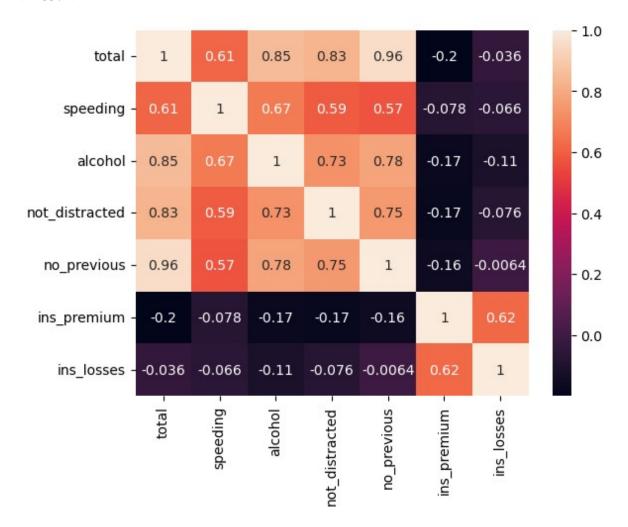
C:\Users\Prasanth Nimmala\AppData\Local\Temp\ipykernel_5936\3182140910.py:1:
FutureWarning: The default value of numeric_only in DataFrame.corr is depreca
ted. In a future version, it will default to False. Select only valid columns
or specify the value of numeric_only to silence this warning.
 corr=df.corr()

Out[13]:

	total	speeding	alcohol	not_distracted	no_previous	ins_premium	ins_lo
total	1.000000	0.611548	0.852613	0.827560	0.956179	-0.199702	-0.03
speeding	0.611548	1.000000	0.669719	0.588010	0.571976	-0.077675	-0.06
alcohol	0.852613	0.669719	1.000000	0.732816	0.783520	-0.170612	-0.112
not_distracted	0.827560	0.588010	0.732816	1.000000	0.747307	-0.174856	-0.07
no_previous	0.956179	0.571976	0.783520	0.747307	1.000000	-0.156895	-0.006
ins_premium	-0.199702	-0.077675	-0.170612	-0.174856	-0.156895	1.000000	0.62
ins_losses	-0.036011	-0.065928	-0.112547	-0.075970	-0.006359	0.623116	1.000

In [14]: sns.heatmap(corr,annot=True)

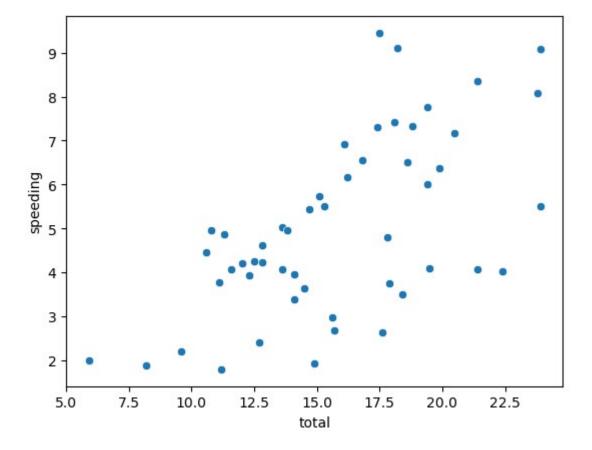
Out[14]: <Axes: >



Scatter Plot

```
In [15]: sns.scatterplot(x='total',y='speeding',data=df)
```

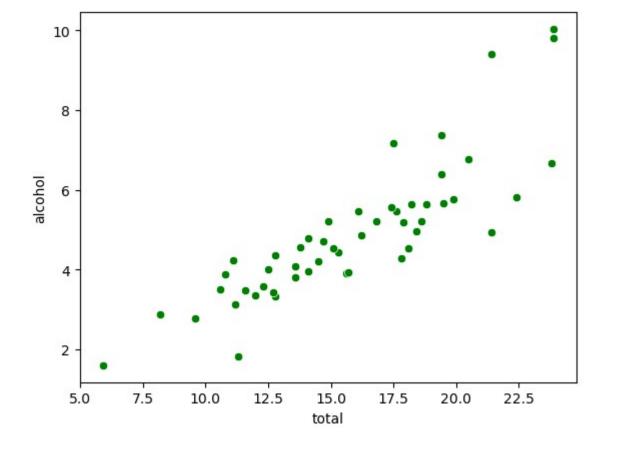
Out[15]: <Axes: xlabel='total', ylabel='speeding'>



In []:

```
In [16]: sns.scatterplot(x='total',y='alcohol',data=df,color='g')
```

Out[16]: <Axes: xlabel='total', ylabel='alcohol'>



In []:

```
In [17]: sns.scatterplot(x='total',y='not_distracted',data=df,color='r')
Out[17]: <Axes: xlabel='total', ylabel='not_distracted'>
              20
           not_distracted
              15
              10
                5
                         7.5
                                  10.0
                                           12.5
                                                    15.0
                                                            17.5
                                                                     20.0
                                                                              22.5
                 5.0
                                                   total
In [ ]:
```

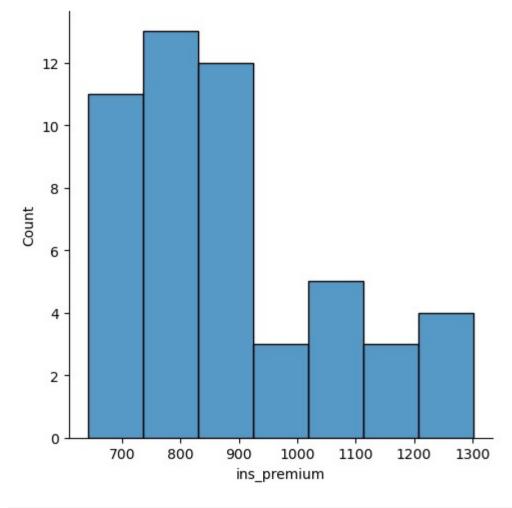
Line Plot

```
In [18]: sns.lineplot(x='total',y='no_previous',data=df,errorbar=None)
Out[18]: <Axes: xlabel='total', ylabel='no_previous'>
              22
              20
              18
              16
           no previous
              10
               8
               6
                         7.5
                                 10.0
                5.0
                                          12.5
                                                   15.0
                                                            17.5
                                                                    20.0
                                                                             22.5
                                                   total
 In [ ]:
```

Distribution Plot

```
In [21]: sns.displot(df['ins_premium'])
```

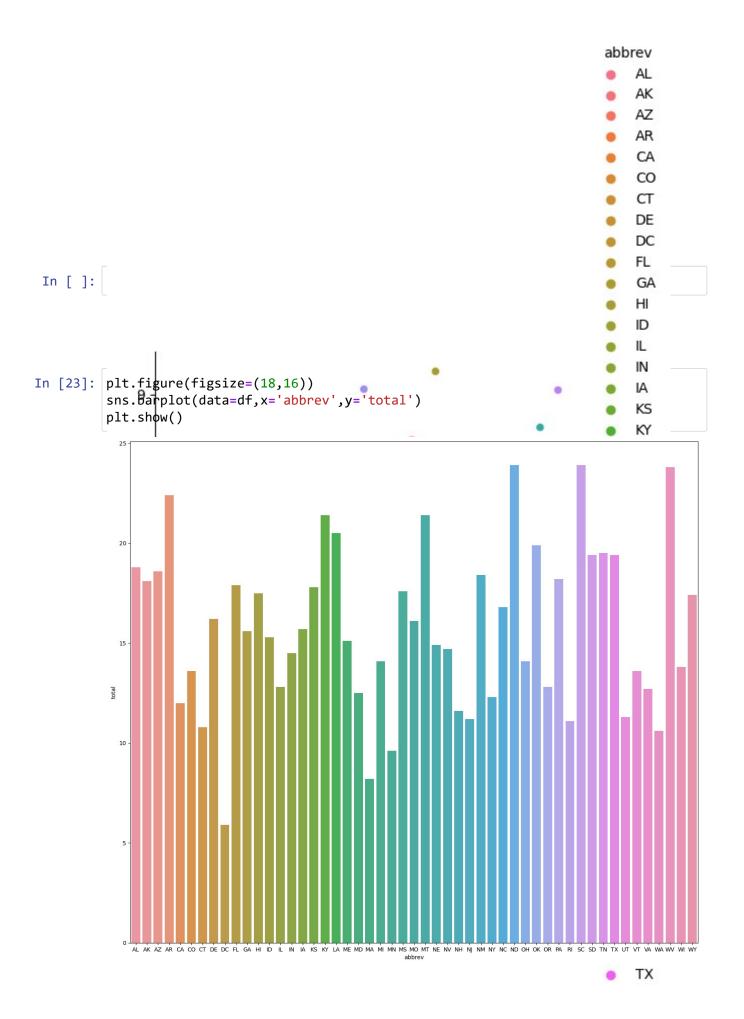
Out[21]: <seaborn.axisgrid.FacetGrid at 0x1c44a90df30>



```
In [ ]:
```

Rel Plot

Out[22]: <seaborn.axisgrid.FacetGrid at 0x1c44aa03130>

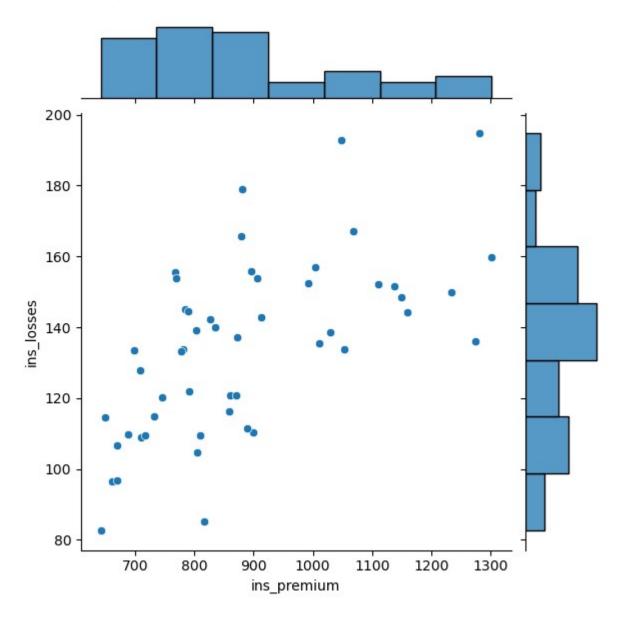


```
In [ ]:
```

Joint Plot

```
In [25]: sns.jointplot(x='ins_premium',y='ins_losses',data=df)
```

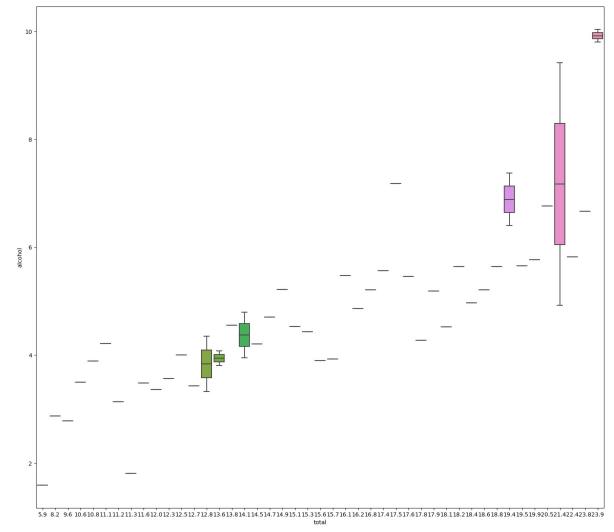
Out[25]: <seaborn.axisgrid.JointGrid at 0x1c44e91f550>



In []:

Box Plot

```
In [26]: plt.figure(figsize=(18,16))
sns.boxplot(x=df['total'],y=df['alcohol'],data=df)
plt.show()
```



```
In [ ]:
```