

Assignment 2

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21BIT0725

```
In [ ]: import seaborn as sns
```

```
In [22]: import numpy as np
```

```
In [23]: import matplotlib.pyplot as plt
```

```
In [28]: dataset=sns.load_dataset('car_crashes')
```

```
In [29]: dataset
```

Out[29]:

	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	MO
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

	total	speeding	alcohol	not_distracted	no_previous	ins_premium	ins_losses	abbrev
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
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	OH
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	13.8	4.968	4.554	5.382	11.592	670.31	106.62	WI
50	17.4	7.308	5.568	14.094	15.660	791.14	122.04	WY

In [30]: `dataset.describe()`

Out[30]:

	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	134.493176
std	4.122002	2.017747	1.729133	4.508977	3.764672	178.296285	24.835938
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 [33]: `print(dataset.head())`

```

      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

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

```

In [31]: `dataset.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   not_distracted         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
dtypes: float64(7), object(1)
memory usage: 3.3+ KB

```

LABEL/COLUMN DEFINITION

total: This column represents the total number of car crashes in a given state.

speeding: It indicates the number of car crashes in which speeding was a contributing factor.

alcohol: This column represents the number of car crashes in which alcohol or drunk driving was a contributing factor.

not_distracted: It likely represents the number of car crashes where the driver was not distracted while driving.

no_previous: This column may represent the number of car crashes where the driver had no previous car crashes on their record.

ins_premium: It stands for insurance prem be the financial losses incurred by insurance companies due to car crashes in a particular state.

abbrev: This column contains the abbreviations (abbreviated names) of the ium, which is the amount of money paid for car insurance.

ins_losses: This represents the insurance losses, which couldU.S. states.

SCATTER PLOT

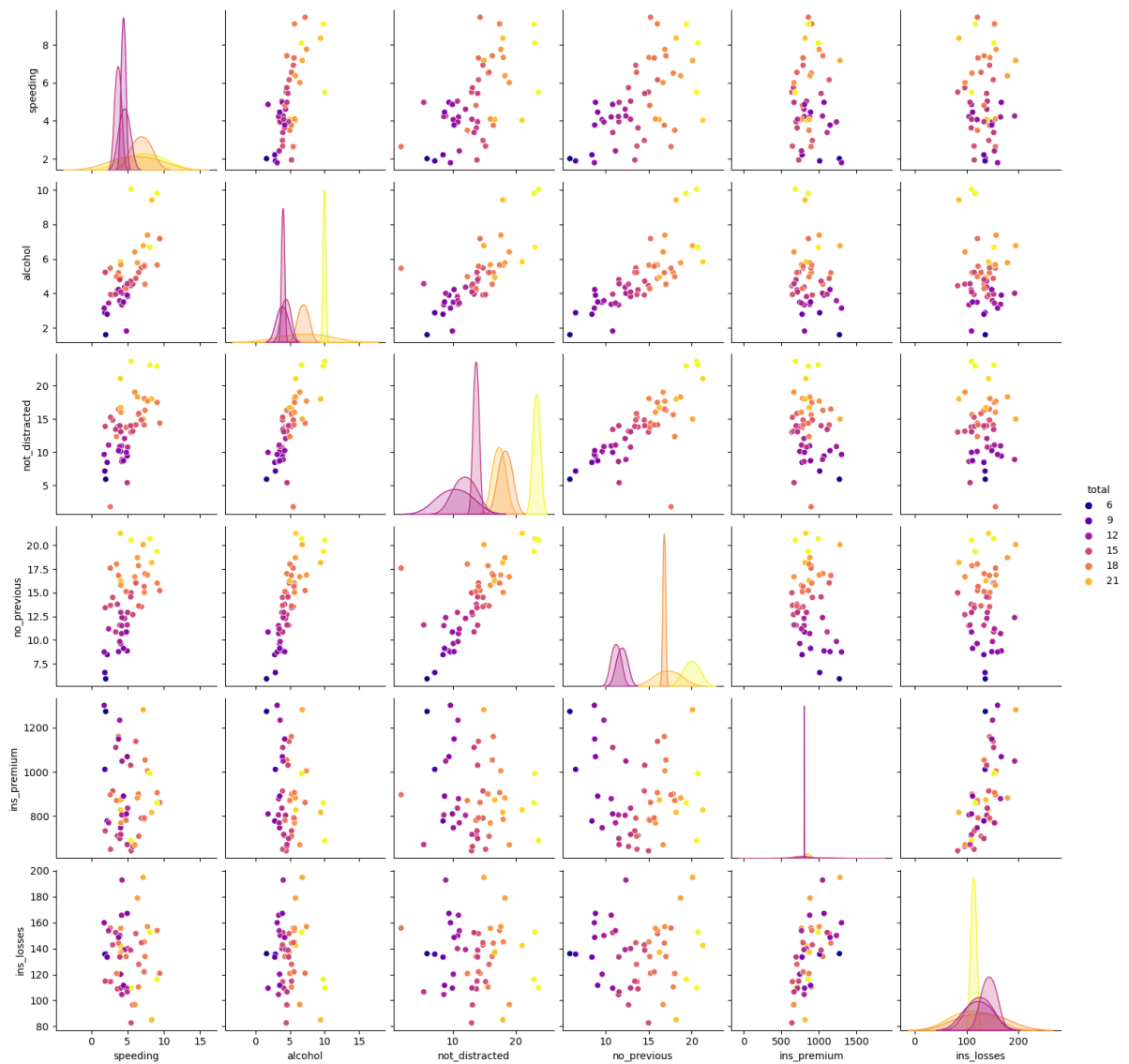
```
In [58]: sns.color_palette("plasma", as_cmap=True)
```

Out[58]: **plasma**



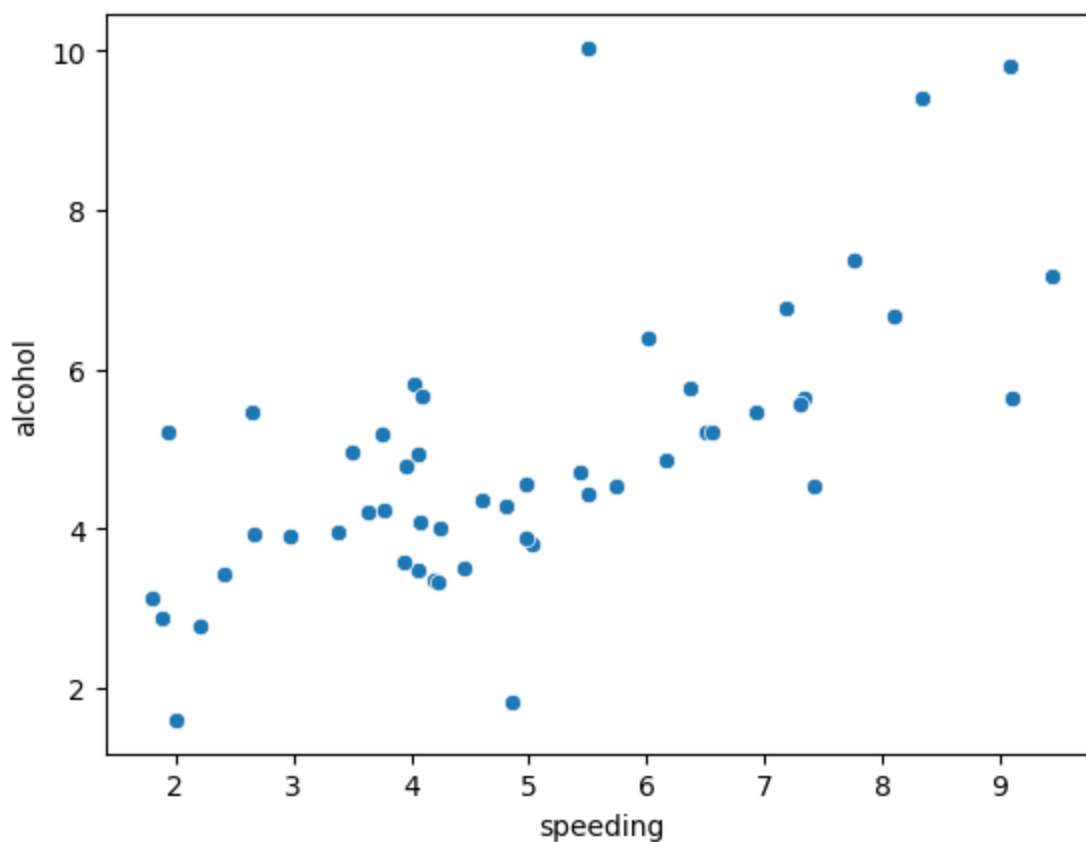
```
In [59]: sns.pairplot(data=dataset, hue="total", kind="scatter", palette="plasma")
```

Out[59]: <seaborn.axisgrid.PairGrid at 0x2cc1da11010>



```
In [65]: sns.scatterplot(data=dataset,x="speeding",y="alcohol")
```

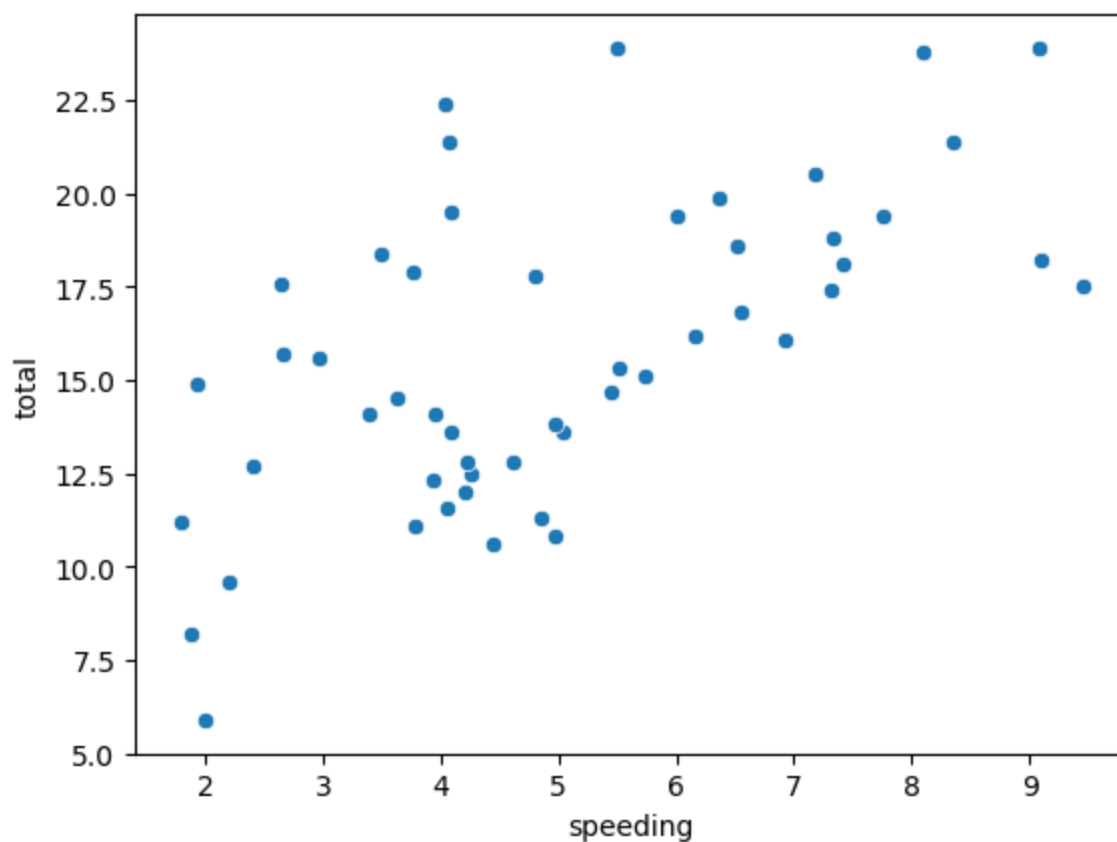
```
Out[65]: <Axes: xlabel='speeding', ylabel='alcohol'>
```



Inference: In the scatter plot, there is a positive correlation between 'alcohol' and 'speeding', implying that states with more alcohol-related crashes also tend to have more speeding-related crashes.

```
In [68]: sns.scatterplot(data=dataset, x="speeding", y="total")
```

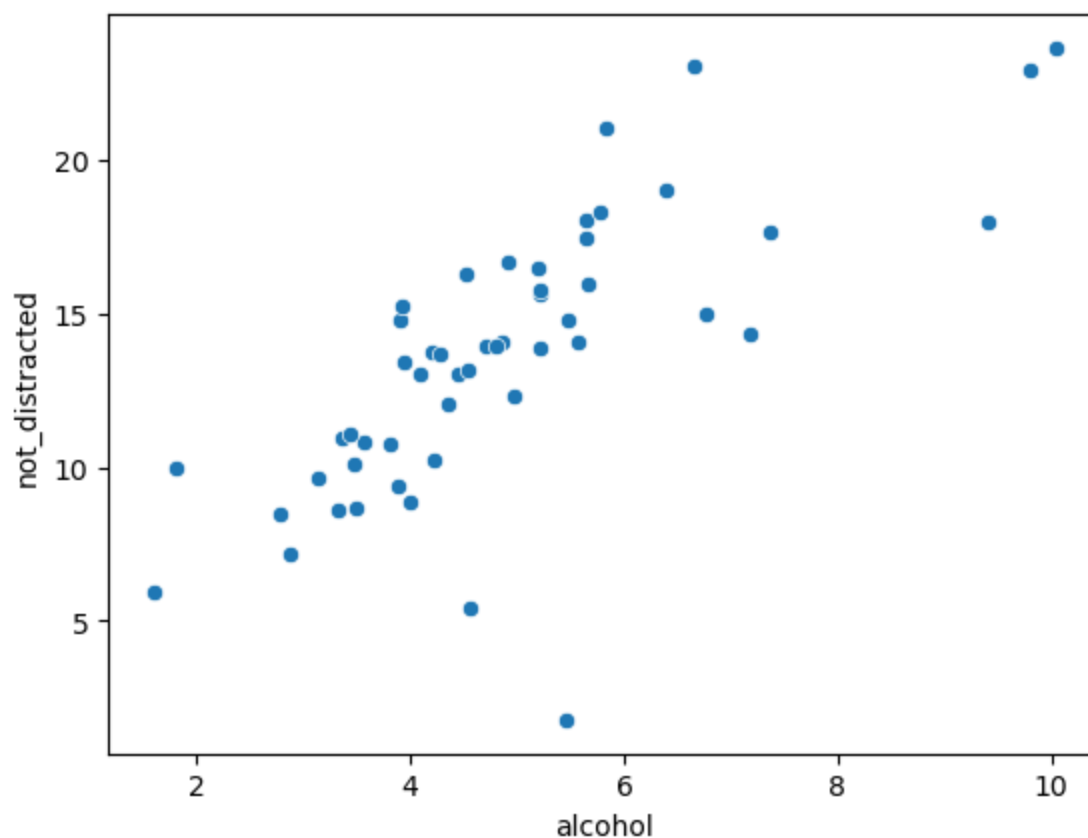
```
Out[68]: <Axes: xlabel='speeding', ylabel='total'>
```



Inference: In the scatter plot comparing 'speeding' and 'total' car crashes, there is no strong linear correlation evident, suggesting that the number of speeding-related crashes does not consistently predict the total number of car crashes across states.

```
In [66]: sns.scatterplot(data=dataset, x="alcohol", y="not_distracted")
```

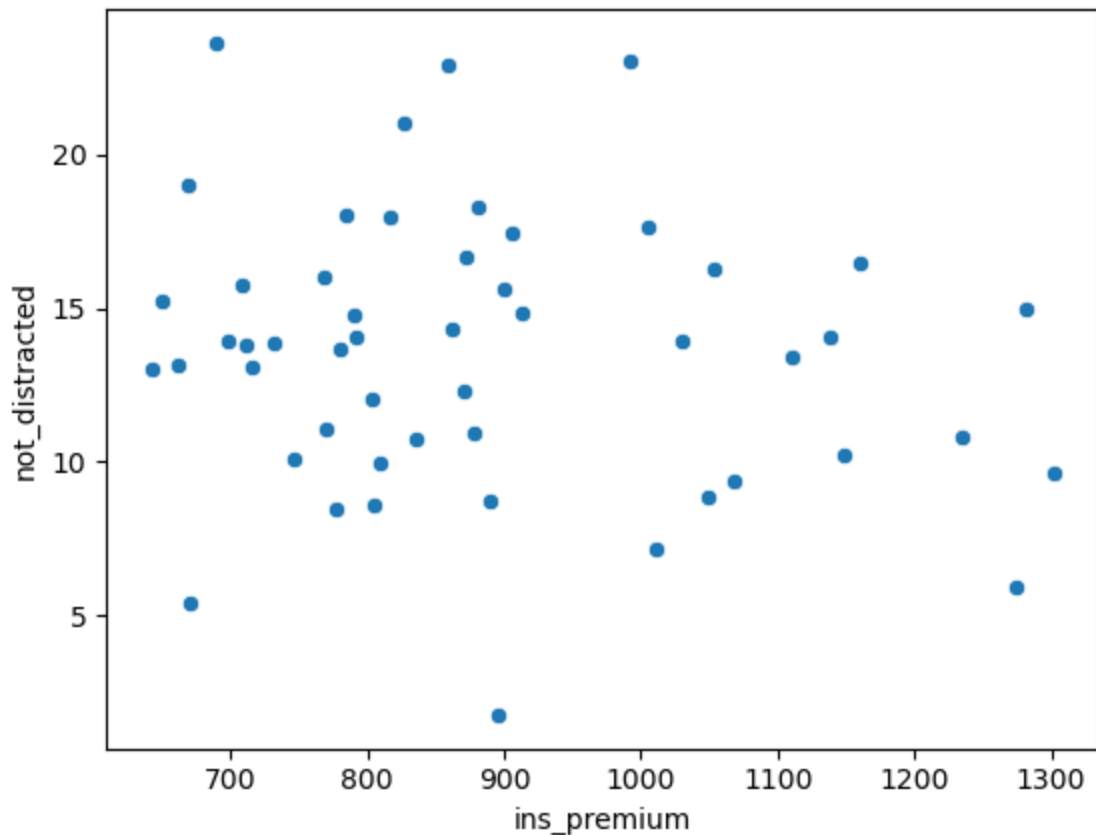
```
Out[66]: <Axes: xlabel='alcohol', ylabel='not_distracted'>
```

Inference: In the scatter plot of 'alcohol' and 'not_distracted' car crashes, there appears to be no strong linear relationship, implying that the presence of alcohol-related crashes does not significantly correlate with cases where drivers were not distracted during accidents

```
In [67]: sns.scatterplot(data=dataset, x="ins_premium", y="not_distracted")
```

```
Out[67]: <Axes: xlabel='ins_premium', ylabel='not_distracted'>
```

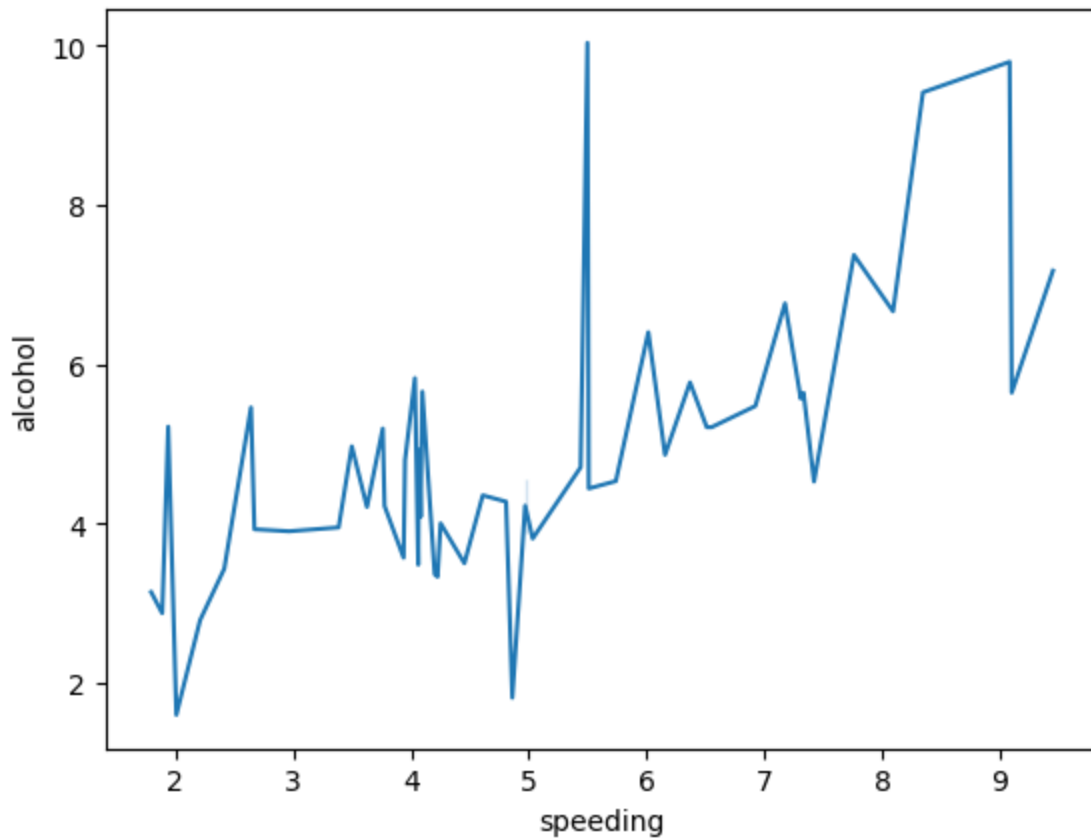


Inference: In the scatter plot of 'ins_premium' and 'not_distracted' car crashes, no clear correlation is evident, suggesting that the cost of insurance premiums does not appear to have a direct relationship with the occurrence of car crashes where drivers were not distracted.

LINE PLOTS

```
In [69]: sns.lineplot(data=dataset, x="speeding", y="alcohol")
```

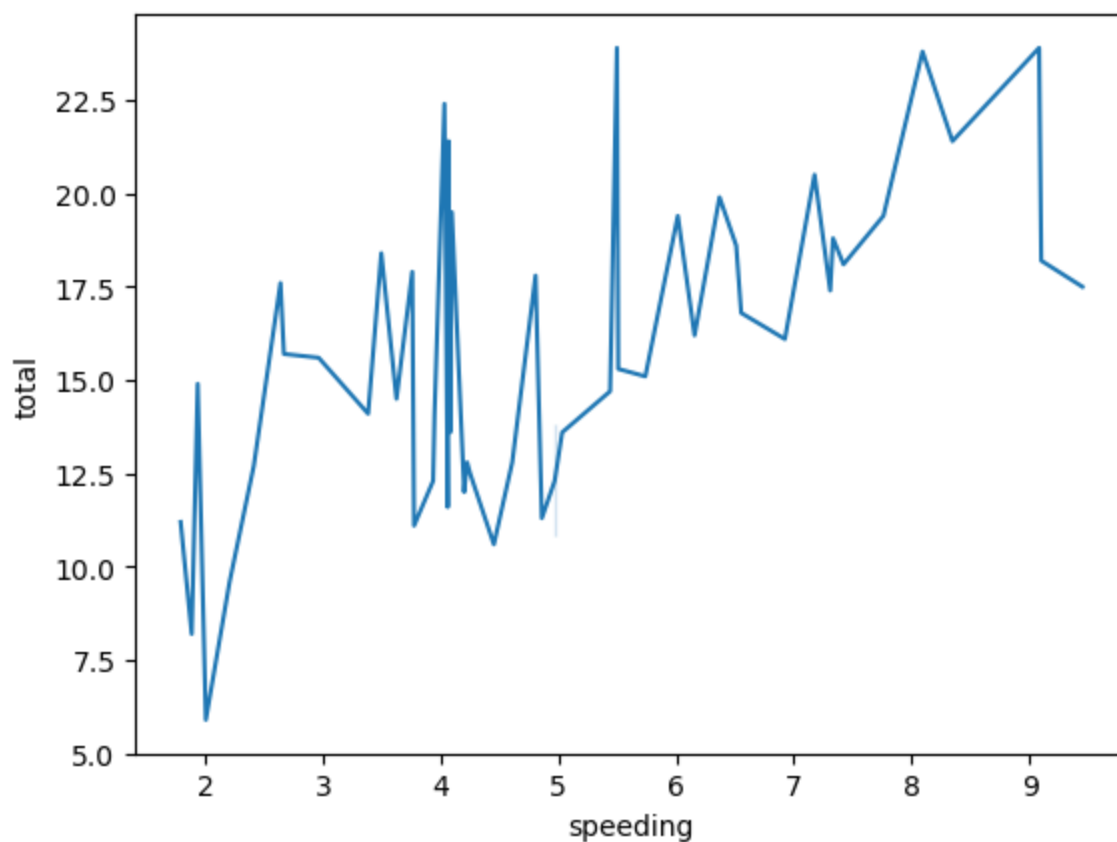
```
Out[69]: <Axes: xlabel='speeding', ylabel='alcohol'>
```



Inference: In the line plot comparing 'speeding' and 'alcohol' across different states, it becomes evident that some states exhibit consistently higher counts of both speeding-related and alcohol-related car crashes, indicating that these risky driving behaviors often co-occur in those regions.

```
In [70]: sns.lineplot(data=dataset, x="speeding", y="total")
```

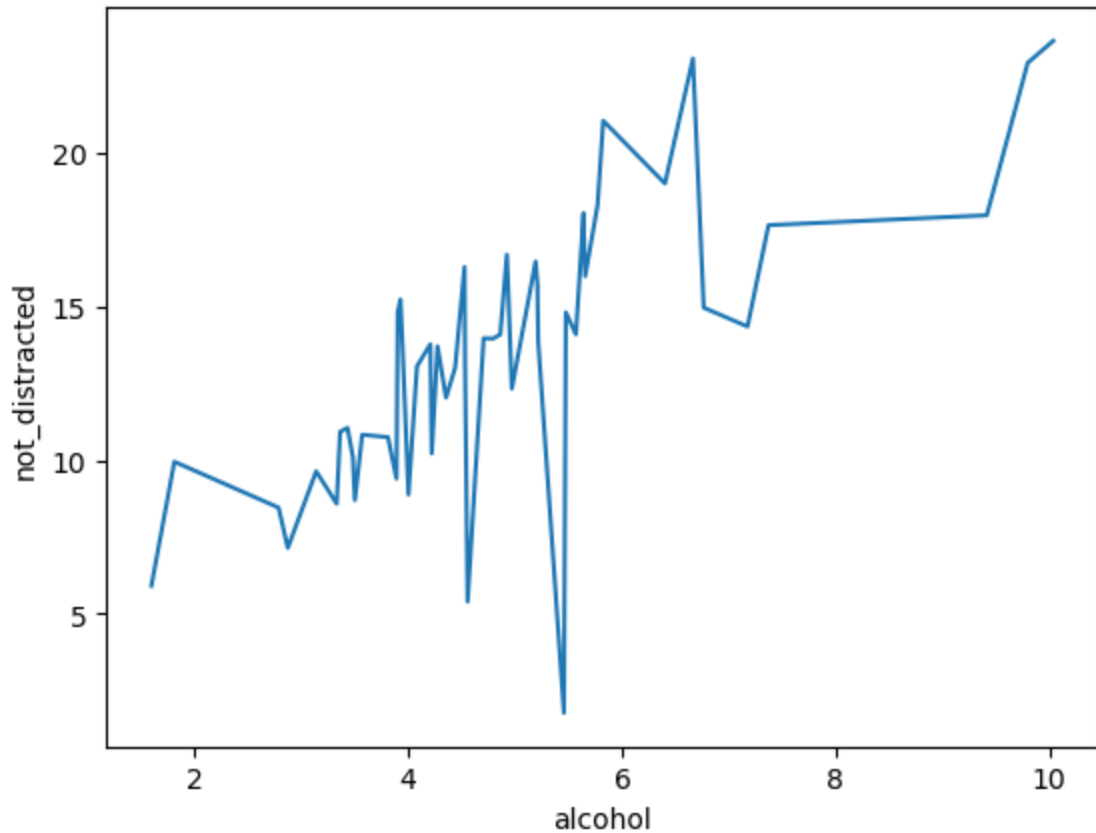
```
Out[70]: <Axes: xlabel='speeding', ylabel='total'>
```



Inference: In the line plot depicting changes in 'speeding' and 'alcohol' car crashes across different states, offering insights into the variability and trends in risky driving behaviors in different regions.

```
In [71]: sns.lineplot(data=dataset, x="alcohol", y="not_distracted")
```

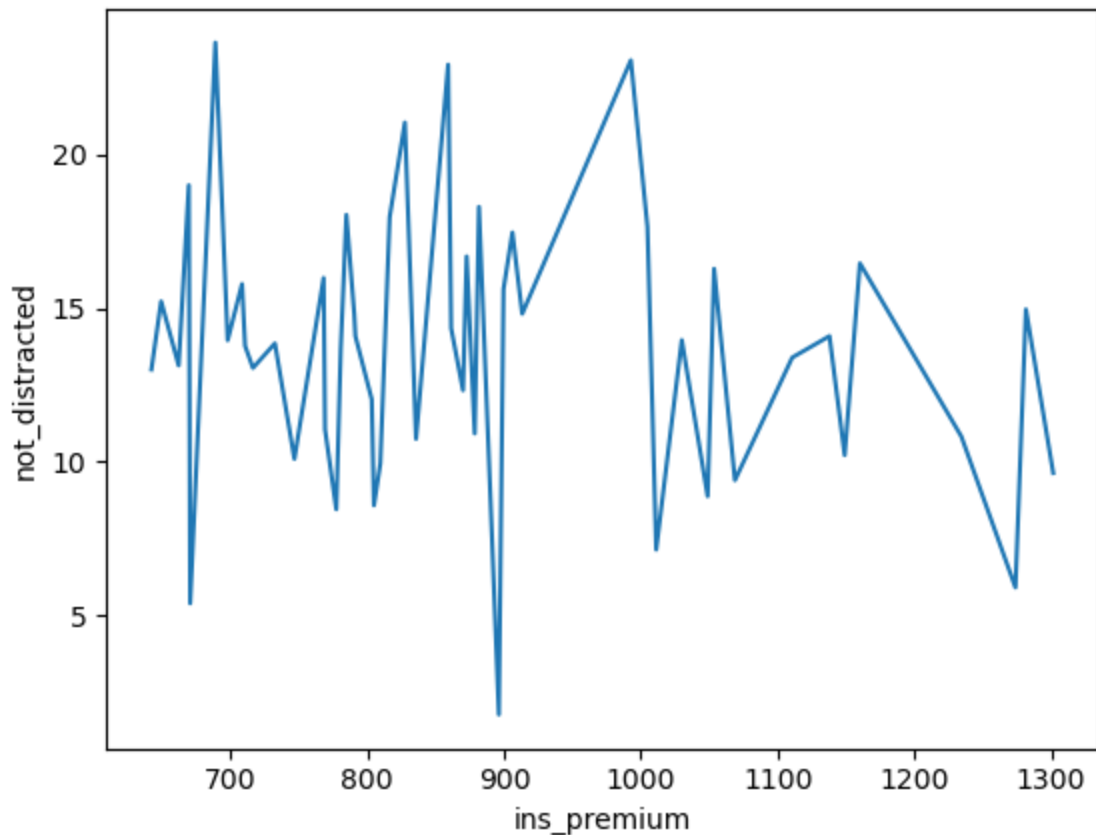
```
Out[71]: <Axes: xlabel='alcohol', ylabel='not_distracted'>
```



Inference: In the line plot illustrating variations in 'alcohol' and 'not_distracted' car crashes across different states, providing insights into the complex relationship between alcohol-related accidents and cases where drivers were not distracted

```
In [72]: sns.lineplot(data=dataset, x="ins_premium", y="not_distracted")
```

```
Out[72]: <Axes: xlabel='ins_premium', ylabel='not_distracted'>
```

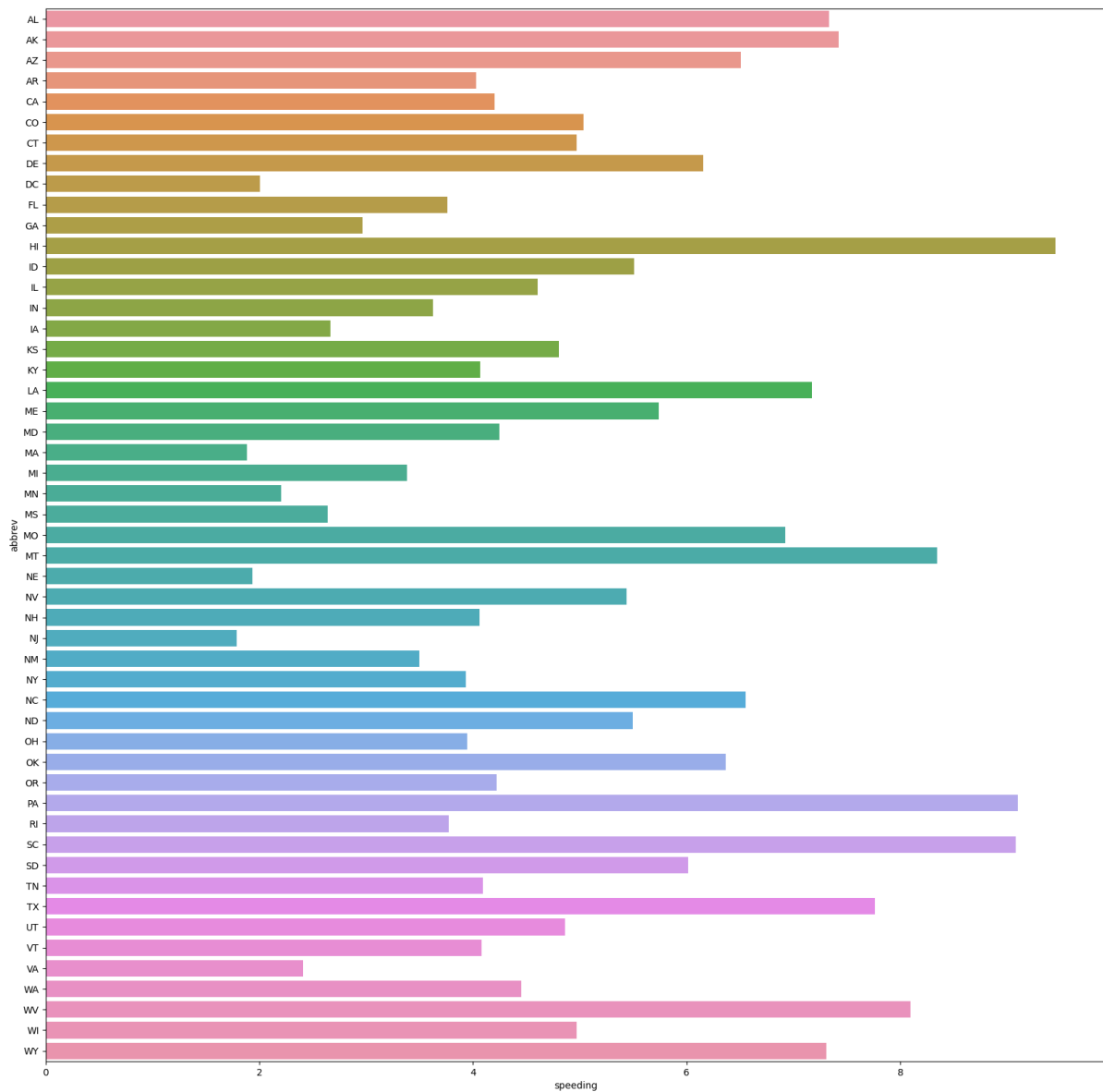


Inference: In the line plot depicting changes in 'ins_premium' and 'not_distracted' car crashes across different states, there seems to be no discernible trend or strong correlation, indicating that the cost of insurance premiums does not appear to be closely related to the occurrence of car crashes where drivers were not distracted

BAR PLOTS

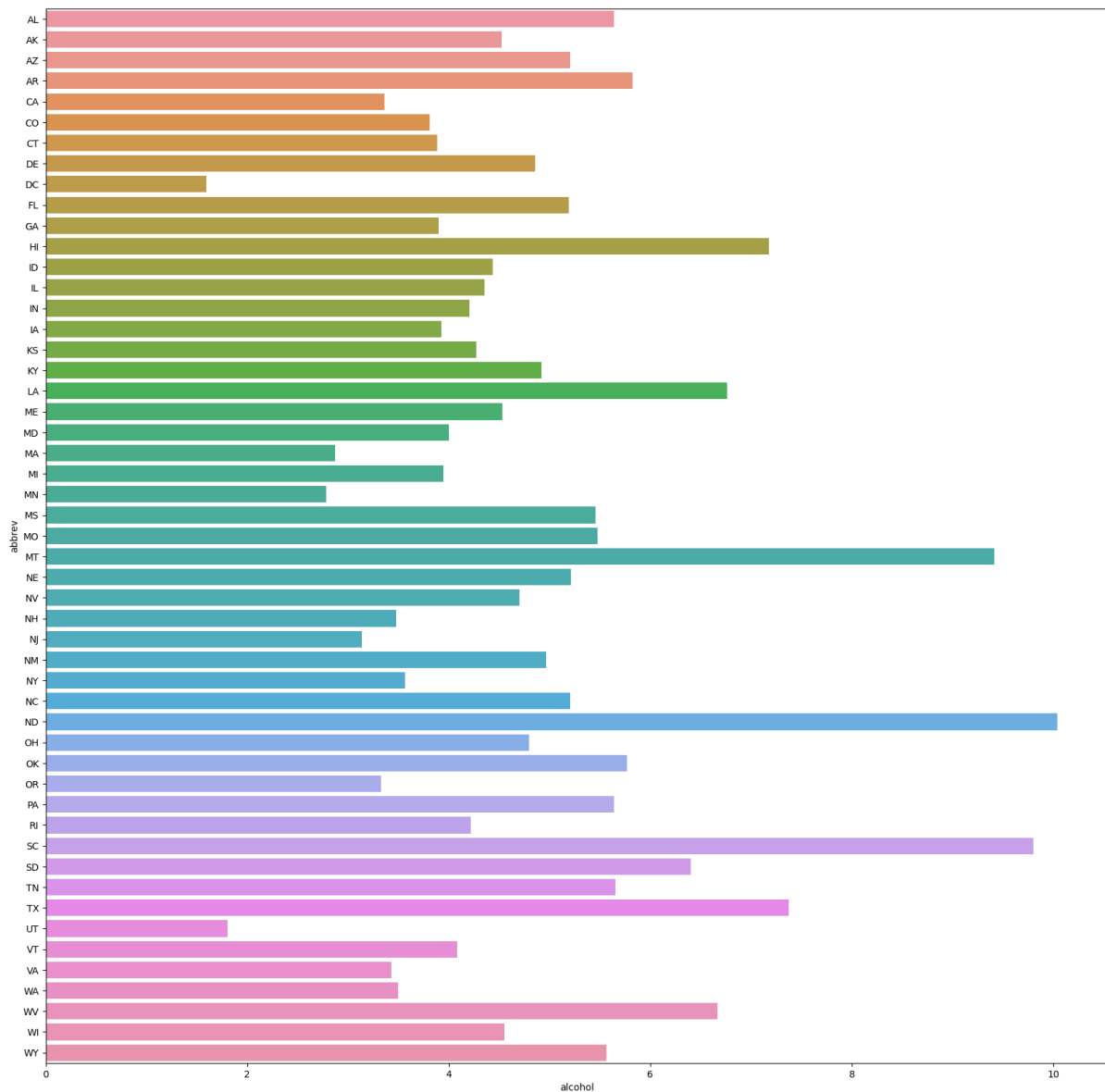
```
In [76]: plt.subplots(figsize=(20,20))  
sns.barplot(data=dataset,x="speeding",y="abbrev")
```

```
Out[76]: <Axes: xlabel='speeding', ylabel='abbrev'>
```



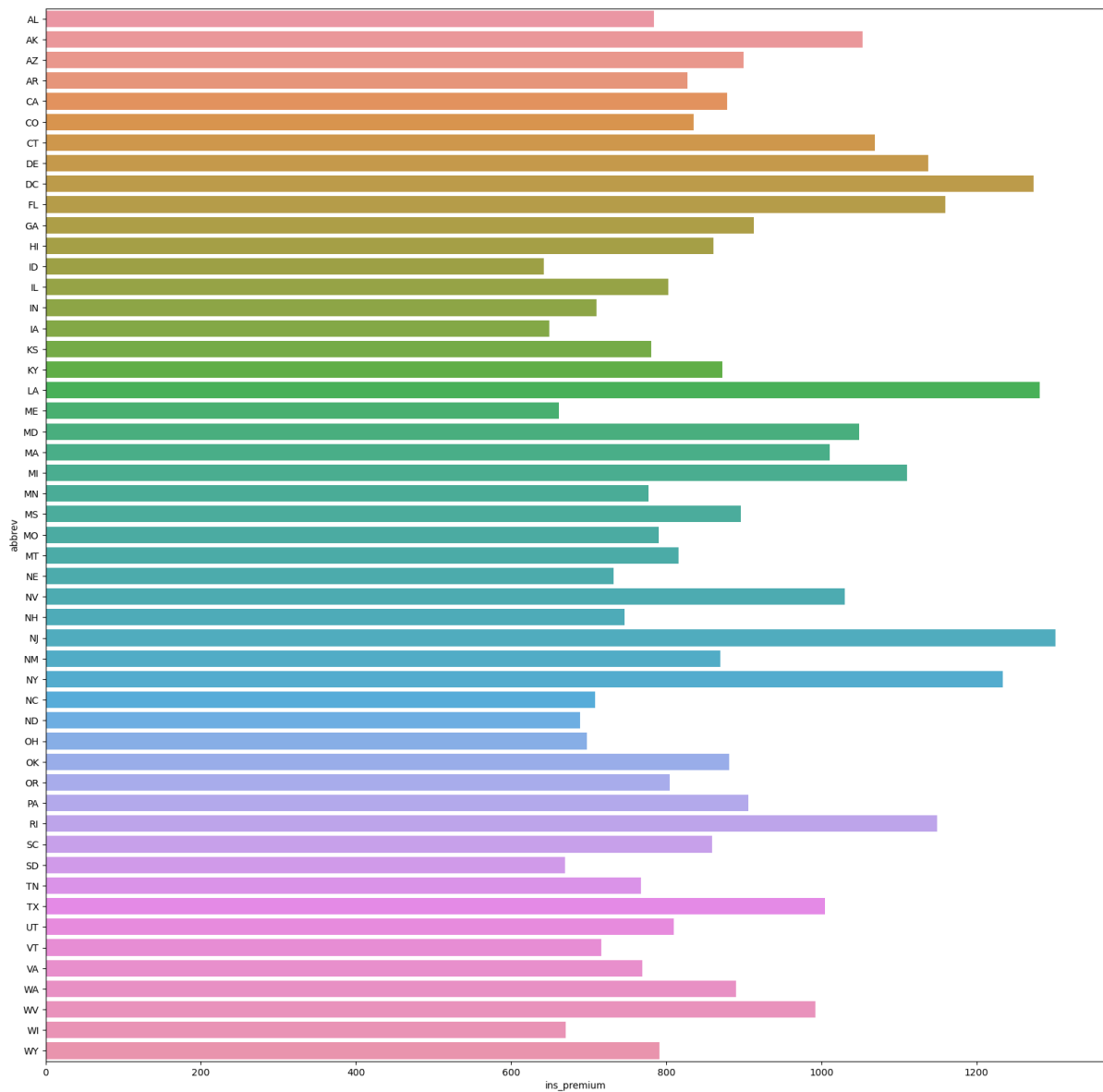
```
In [77]: plt.subplots(figsize=(20,20))  
sns.barplot(data=dataset,x="alcohol",y="abbrev")
```

```
Out[77]: <Axes: xlabel='alcohol', ylabel='abbrev'>
```



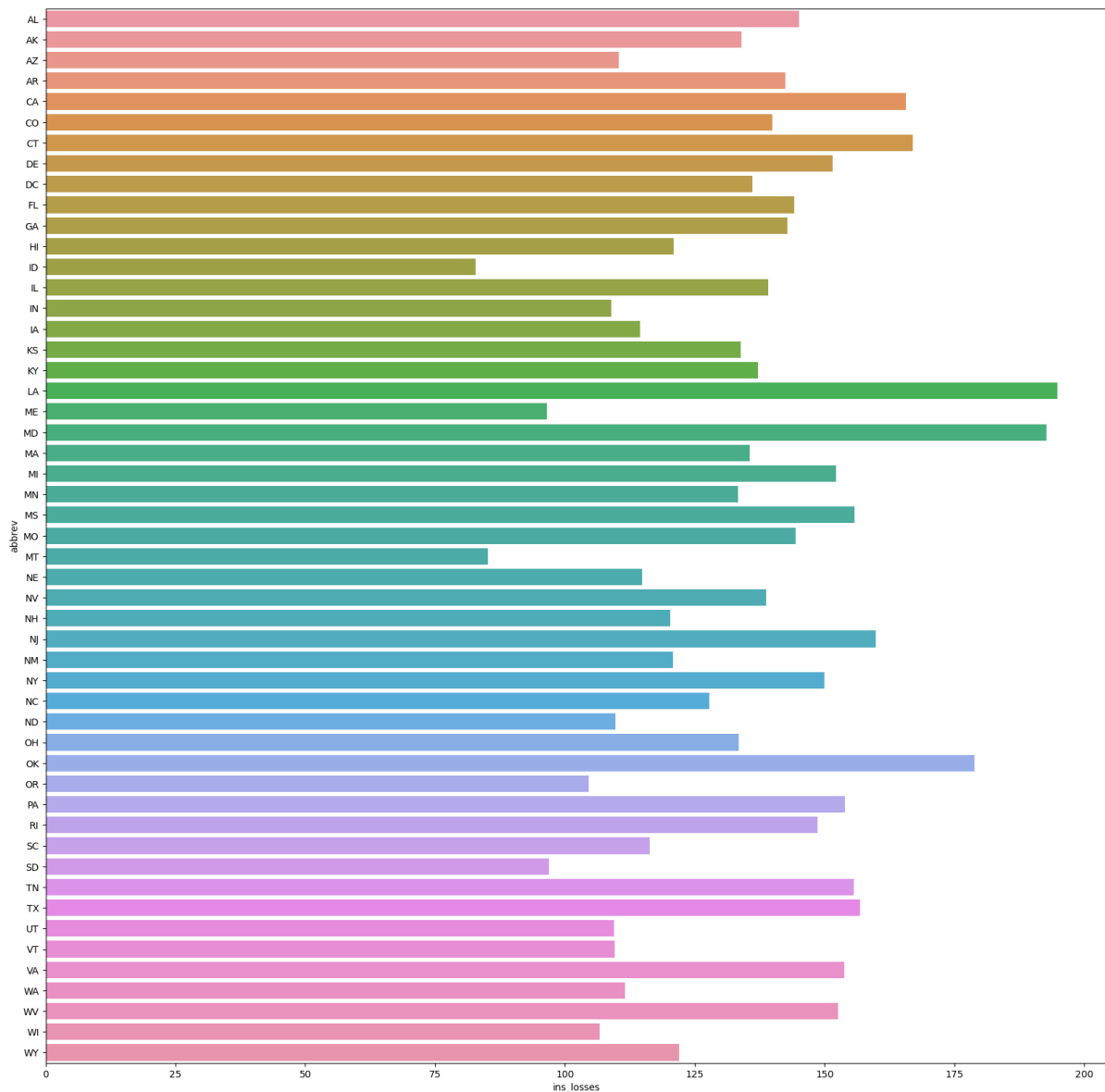
```
In [79]: plt.subplots(figsize=(20,20))
sns.barplot(data=dataset,x="ins_premium",y="abbrev")
```

```
Out[79]: <Axes: xlabel='ins_premium', ylabel='abbrev'>
```

```
In [80]: plt.subplots(figsize=(20,20))
sns.barplot(data=dataset,x="ins_losses",y="abbrev")
```

```
Out[80]: <Axes: xlabel='ins_losses', ylabel='abbrev'>
```



HEAT MAP

```
In [83]: corr = dataset.corr()
         corr
```

C:\Users\WELCOME\AppData\Local\Temp\ipykernel_14272\897440734.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 = dataset.corr()
```

Out[83]:

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

In [85]: `sns.heatmap(corr,annot=True,cmap="magma")`Out[85]: `<Axes: >`