# **Assignment - 2**

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```
In [1]:
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 [3]:
df = sns.load_dataset('car_crashes')
```

In [4]:

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     |
| 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     |

|              | total        | speeding | alcohol | not_distracted | no_previous | ins_premium | ins_losses | abbrev |
|--------------|--------------|----------|---------|----------------|-------------|-------------|------------|--------|
| 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     |
| <b>50</b> In | 17.4<br>[5]: | 7.308    | 5.568   | 14.094         | 15.660      | 791.14      | 122.04     | WY     |

df.head()

Out[5]:

|   | 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     |

In [6]:

df.tail()

Out[6]:

|    | total | speeding | alcohol | not_distracted | no_previous | ins_premium | ins_losses | abbrev |
|----|-------|----------|---------|----------------|-------------|-------------|------------|--------|
| 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 [7]:

```
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 51 non-null 1 speeding float64 2 alcohol 51 non-null float64 3 not\_distracted 51 non-null float64 4 51 non-null float64 no previous 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

#### In [8]:

```
df.shape
```

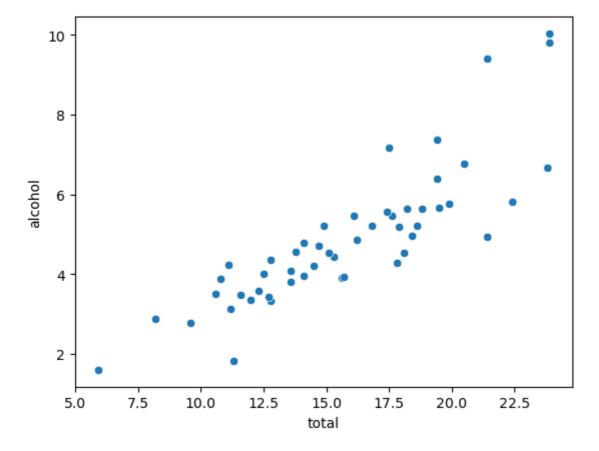
#### Out[8]:

(51, 8)

#### In [9]:

```
sns.scatterplot(x="total" , y ="alcohol", data = df)
print("Alcohol is directly proportional to total")
```

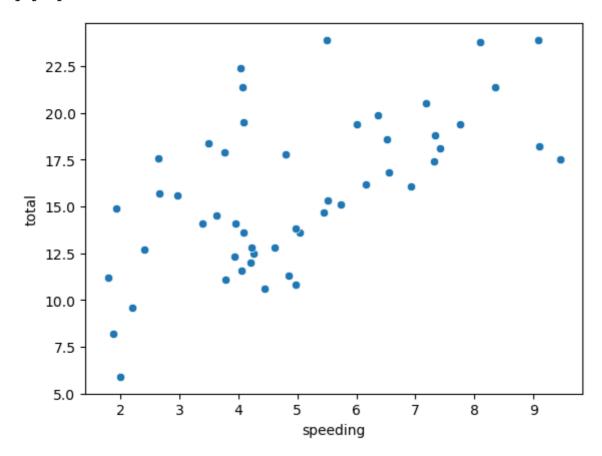
Alcohol is directly proportional to total



#### In [10]:

```
"speeding" , y = "total", data = df)
directly proportional to total whereas speeding is inversely proportional to total")
```

Alcohol is directly proportional to total whereas speeding is inversel y proportional to total

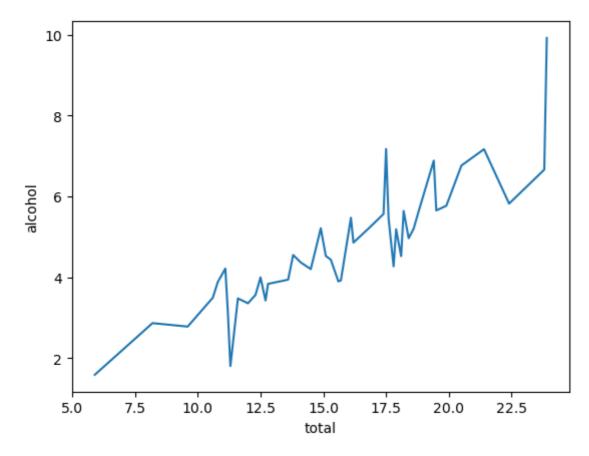


## In [11]:

```
sns.lineplot(x="total" , y ="alcohol", data = df, errorbar= None)
```

## Out[11]:

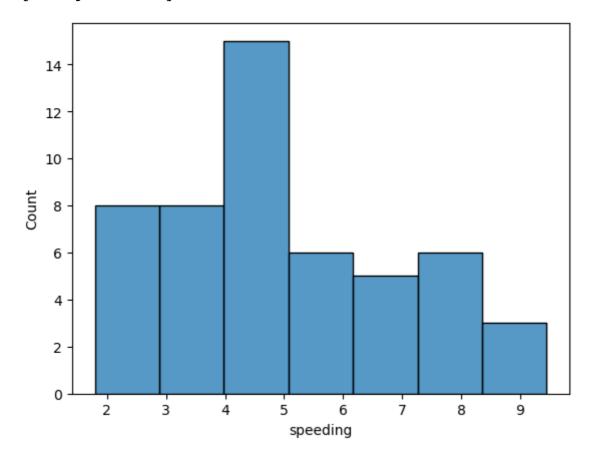
<Axes: xlabel='total', ylabel='alcohol'>



## In [12]:

```
sns.histplot(df["speeding"])
print("Speeding vs Density distribution")
```

# Speeding vs Density distribution



```
In [13]:
```

```
sns.relplot(x="speeding" , y ="alcohol",data =df,hue = "abbrev")
```

## Out[13]:

<seaborn.axisgrid.FacetGrid at 0x14a4e7a90>

## In [14]:

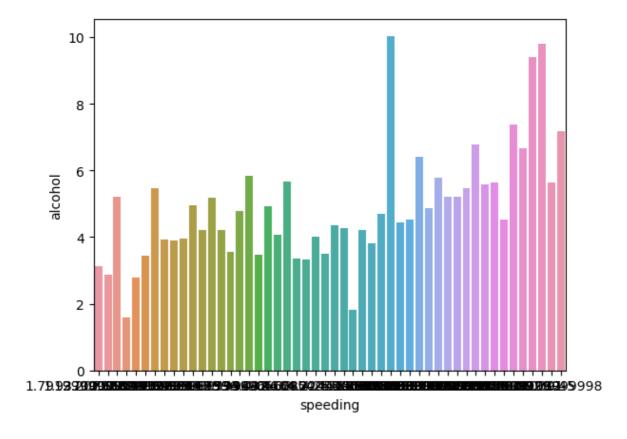
```
df["abbrev"].value_counts()
print("Here, we get count of all categories and has everything repeated")
```

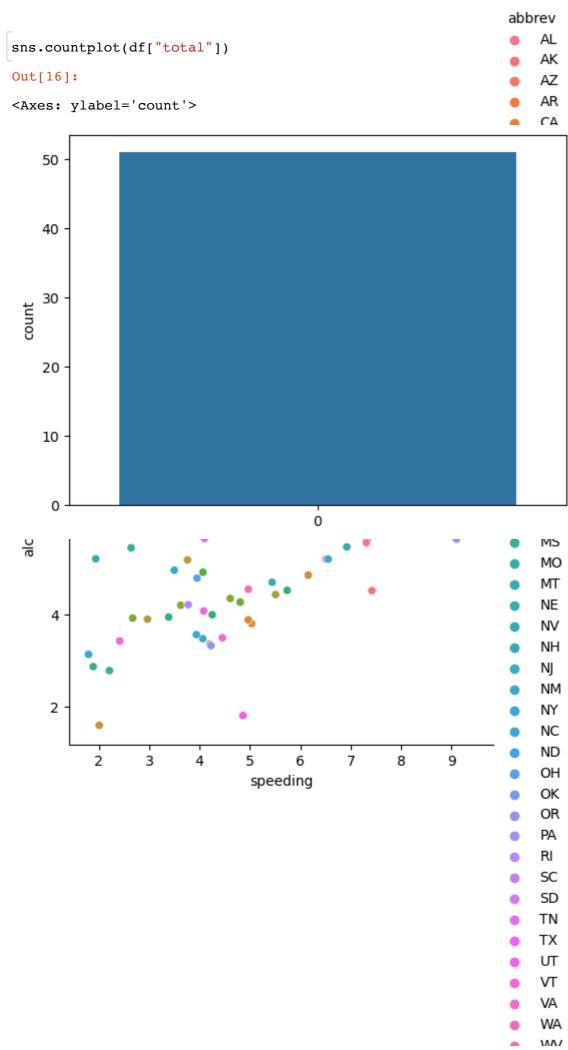
Here, we get count of all categories and has everything repeated

#### In [15]:

```
sns.barplot(data = df,x = "speeding",y = "alcohol",errorbar = None)
print("bargraph: speeding vs alcohol")
```

bargraph: speeding vs alcohol

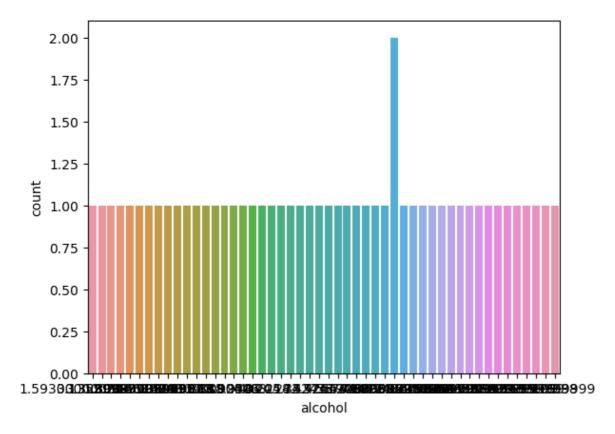




sns.countplot(data= df,x = "alcohol")

## Out[17]:

<Axes: xlabel='alcohol', ylabel='count'>

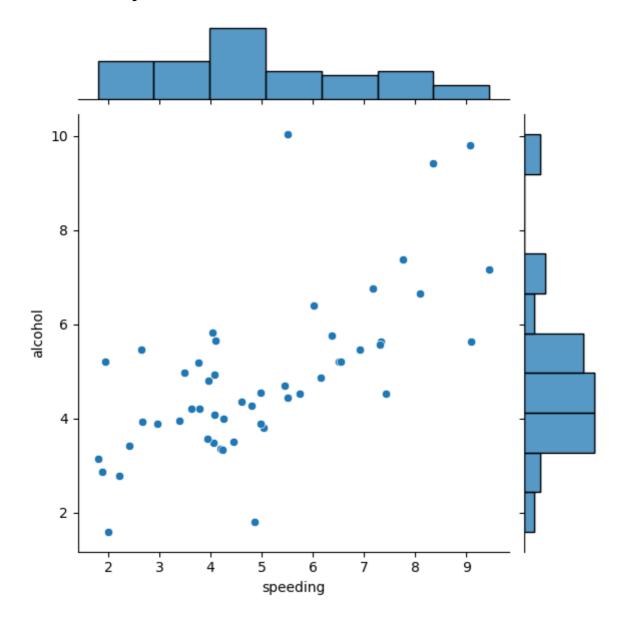


# In [18]:

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

# Out[18]:

<seaborn.axisgrid.JointGrid at 0x14a71ea10>

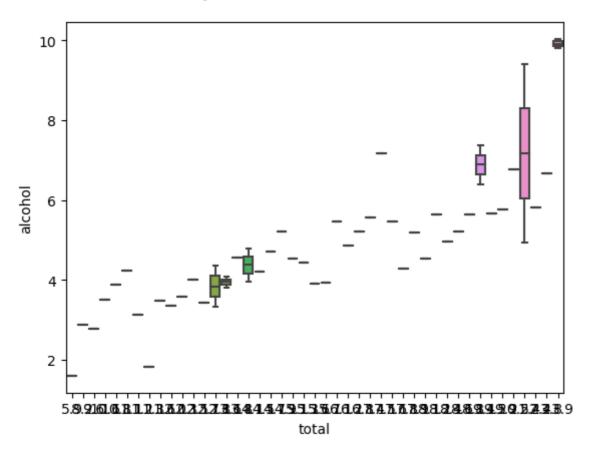


#### In [19]:

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

#### Out[19]:

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



#### In [20]:

```
corr= df.corr()
corr
```

/var/folders/0g/xqmh0yz92jx\_s8ljsv3x08wr0000gn/T/ipykernel\_36494/38645 96578.py:1: FutureWarning: The default value of numeric\_only in DataFr ame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to sile nce this warning.

corr= df.corr()

### Out[20]:

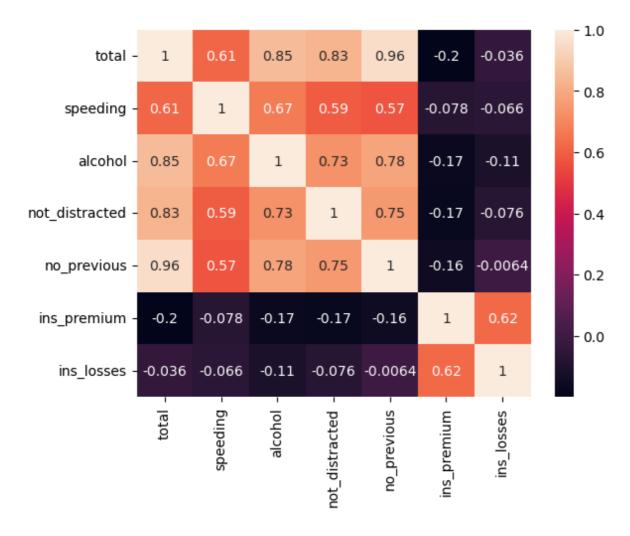
|                | total     | speeding  | alcohol   | not_distracted | no_previous | ins_premium | ins_los |
|----------------|-----------|-----------|-----------|----------------|-------------|-------------|---------|
| total          | 1.000000  | 0.611548  | 0.852613  | 0.827560       | 0.956179    | -0.199702   | -0.036  |
| speeding       | 0.611548  | 1.000000  | 0.669719  | 0.588010       | 0.571976    | -0.077675   | -0.065  |
| 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   | -0.075  |
| 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    | 0.623   |
| ins_losses     | -0.036011 | -0.065928 | -0.112547 | -0.075970      | -0.006359   | 0.623116    | 1.000   |

#### In [21]:

## sns.heatmap(corr,annot= True)

#### Out[21]:

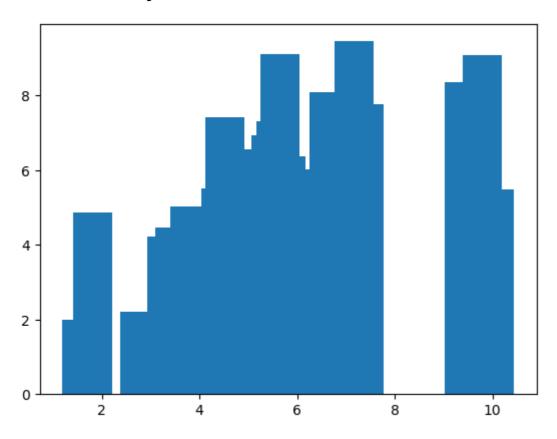
#### <Axes: >



## In [22]:

```
x=df["alcohol"]
y = df["speeding"]
plt.bar(x,y)
```

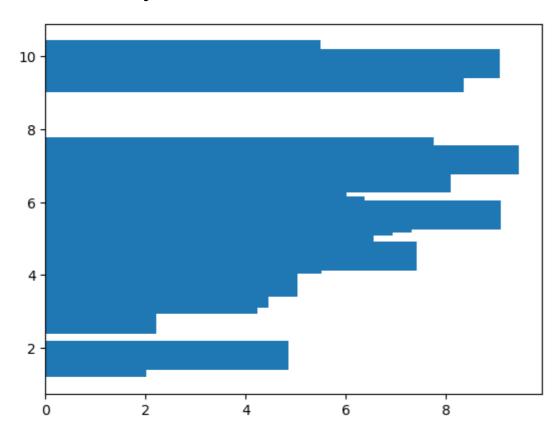
# Out[22]:



# In [23]:

plt.barh(x,y)

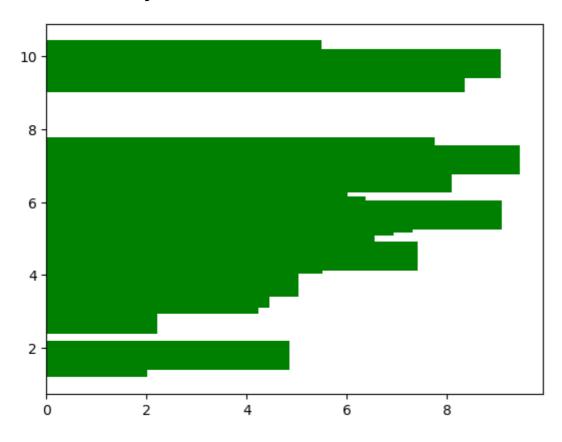
# Out[23]:



# In [24]:

plt.barh(x,y,color = 'green')

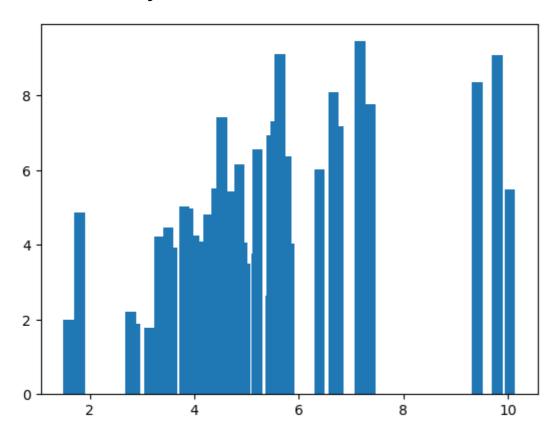
# Out[24]:



# In [25]:

plt.bar(x,y,width = 0.2)

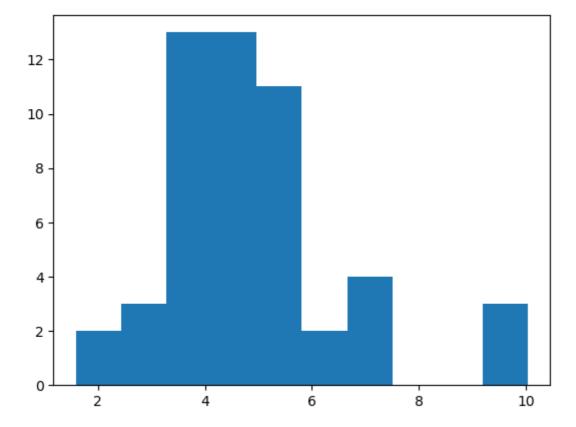
# Out[25]:



#### In [26]:

```
plt.hist(x)
```

## Out[26]:



```
In [27]:
```

```
x1 =(df["alcohol"])
fig = plt.figure()
axes1 = fig.add_axes([0.1,0.1,0.8,0.8])
axes1.pie(x1,y,autopct="%0.2f%%",colors=["red","green"])
Out[27]:
```

```
([<matplotlib.patches.Wedge at 0x14b223690>,
 <matplotlib.patches.Wedge at 0x14b278750>,
 <matplotlib.patches.Wedge at 0x14b27a390>,
 <matplotlib.patches.Wedge at 0x14b279a50>,
 <matplotlib.patches.Wedge at 0x14b281c10>,
 <matplotlib.patches.Wedge at 0x14b283910>,
 <matplotlib.patches.Wedge at 0x14b295550>,
 <matplotlib.patches.Wedge at 0x14b297150>,
 <matplotlib.patches.Wedge at 0x14b2a0d90>,
 <matplotlib.patches.Wedge at 0x14b283790>,
 <matplotlib.patches.Wedge at 0x14b2b0550>,
 <matplotlib.patches.Wedge at 0x14b2b2090>,
 <matplotlib.patches.Wedge at 0x14b2b3d10>,
 <matplotlib.patches.Wedge at 0x14b2bda50>,
 <matplotlib.patches.Wedge at 0x14b2bf650>,
 <matplotlib.patches.Wedge at 0x14b2c9310>,
 <matplotlib.patches.Wedge at 0x14b2caf90>,
 <matplotlib.patches.Wedge at 0x14b2d4d90>.
```

#### In [ ]: