

# seaborn

July 7, 2024

Is a data visualization library

```
[1]: import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
```

Importing dataset directly from seaborn

total\_bill vs tip data set

```
[2]: tips = sns.load_dataset('tips')
```

```
[3]: tips.head()
```

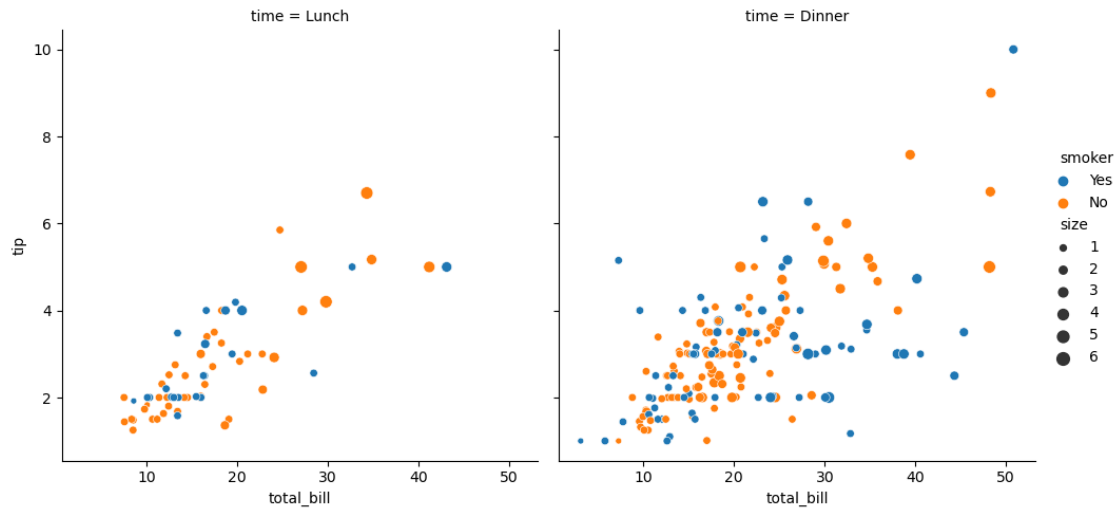
```
[3]:   total_bill  tip  sex smoker  day  time  size
0      16.99  1.01 Female    No  Sun  Dinner     2
1      10.34  1.66  Male    No  Sun  Dinner     3
2      21.01  3.50  Male    No  Sun  Dinner     3
3      23.68  3.31  Male    No  Sun  Dinner     2
4      24.59  3.61 Female    No  Sun  Dinner     4
```

```
[4]: sns.relplot(data=tips, x='total_bill', y='tip', col='time', hue='smoker',
↪size='size');
```

C:\Users\DONATUS\anaconda24\Lib\site-packages\seaborn\axisgrid.py:118:

UserWarning: The figure layout has changed to tight

self.\_figure.tight\_layout(\*args, \*\*kwargs)



Setting Teams for the plot

```
[5]: sns.set_theme()
```

```
[6]: sns.relplot(data=tips, x='total_bill', y='tip', col='time', hue='smoker',
    ↪size='size');
```

C:\Users\DONATUS\anaconda24\Lib\site-packages\seaborn\axisgrid.py:118:  
UserWarning: The figure layout has changed to tight  
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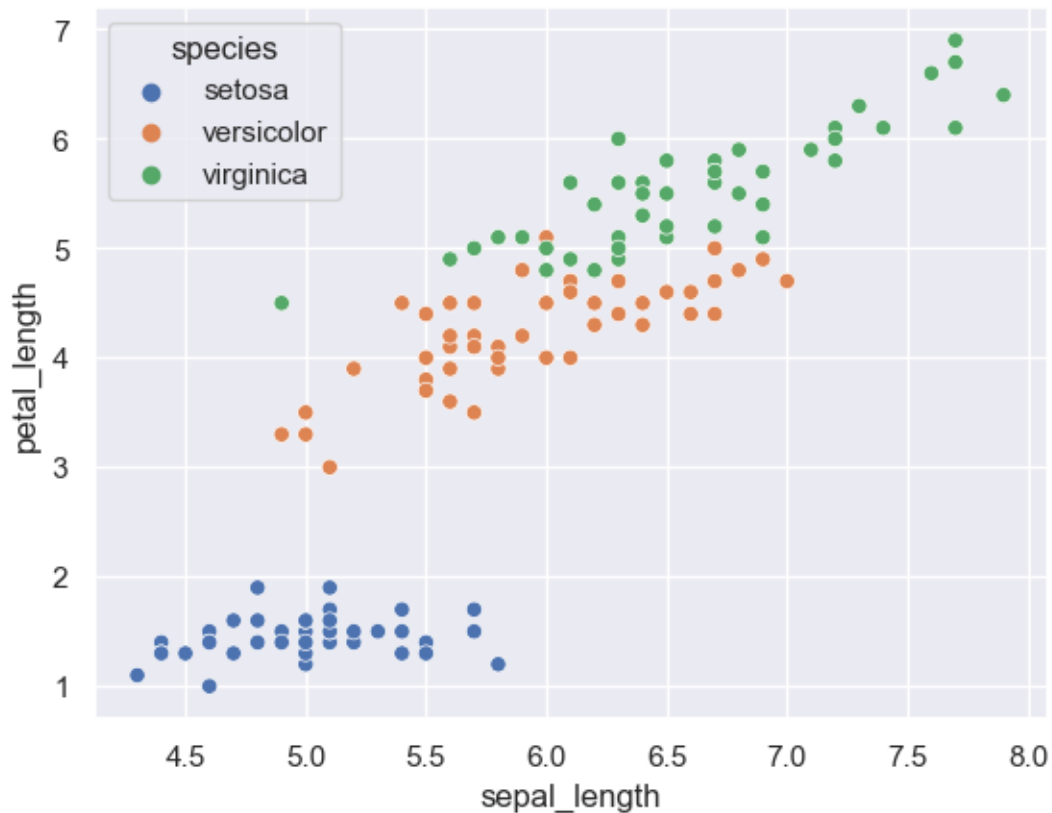
```
[7]: #loadidng iris data set
```

```
iris = sns.load_dataset('iris')
iris.head()
```

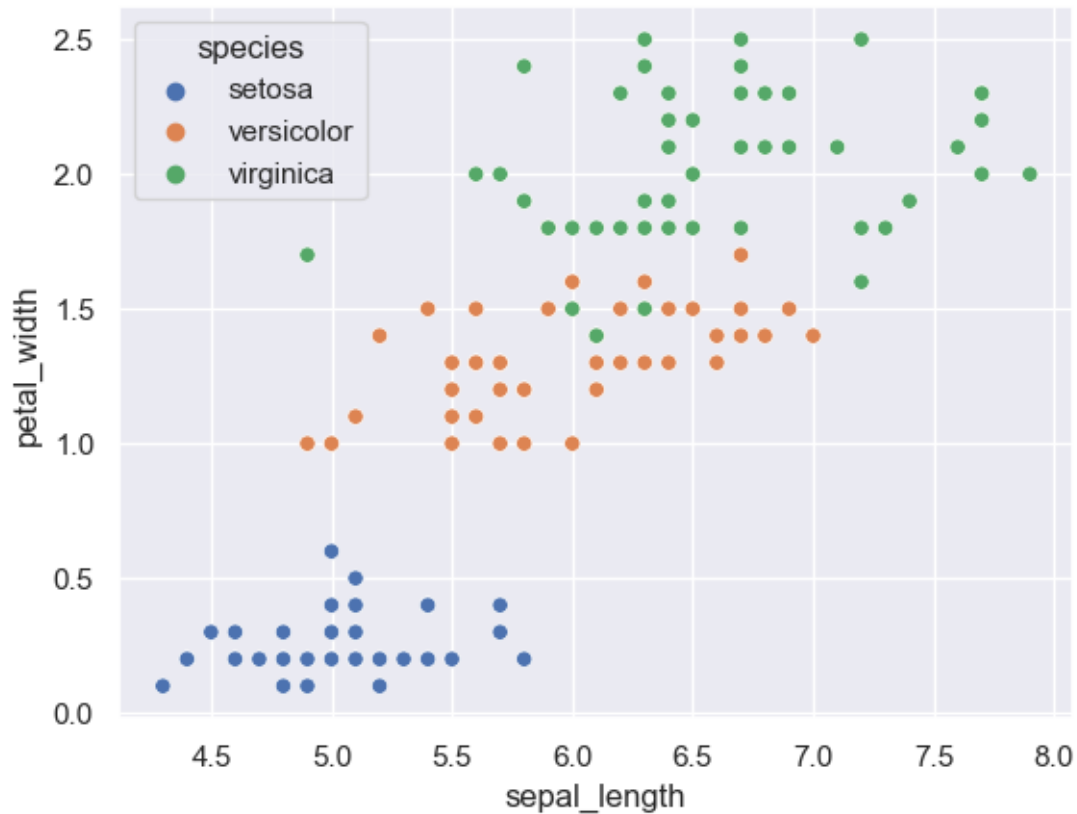
```
[7]:   sepal_length  sepal_width  petal_length  petal_width  species
0          5.1         3.5         1.4         0.2   setosa
1          4.9         3.0         1.4         0.2   setosa
2          4.7         3.2         1.3         0.2   setosa
3          4.6         3.1         1.5         0.2   setosa
4          5.0         3.6         1.4         0.2   setosa
```

Using a scatterplot

```
[8]: sns.scatterplot(x='sepal_length', y='petal_length', hue = 'species', data = iris);
```



```
[9]: sns.scatterplot(x='sepal_length', y='petal_width', hue = 'species', data = iris);
```



```
[10]: #loading the titanic dataset

titanic = sns.load_dataset('titanic')
titanic.head()
```

```
[10]:   survived  pclass    sex  age  sibsp  parch    fare embarked  class \
0         0      3   male  22.0     1     0   7.2500         S  Third
1         1      1  female  38.0     1     0  71.2833         C  First
2         1      3  female  26.0     0     0   7.9250         S  Third
3         1      1  female  35.0     1     0  53.1000         S  First
4         0      3   male  35.0     0     0   8.0500         S  Third

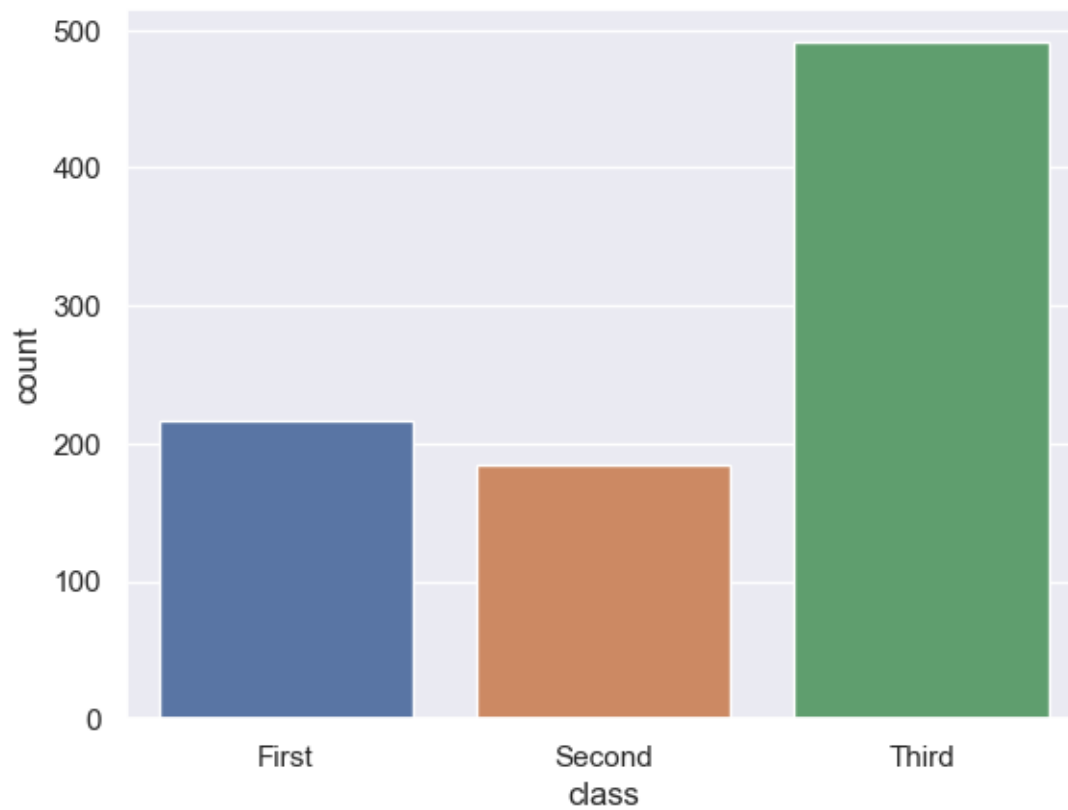
      who  adult_male  deck  embark_town  alive  alone
0    man         True  NaN  Southampton    no  False
1  woman        False   C   Cherbourg   yes  False
2  woman        False  NaN  Southampton   yes   True
3  woman        False   C   Southampton   yes  False
4    man         True  NaN  Southampton    no   True
```

```
[11]: titanic.shape
```

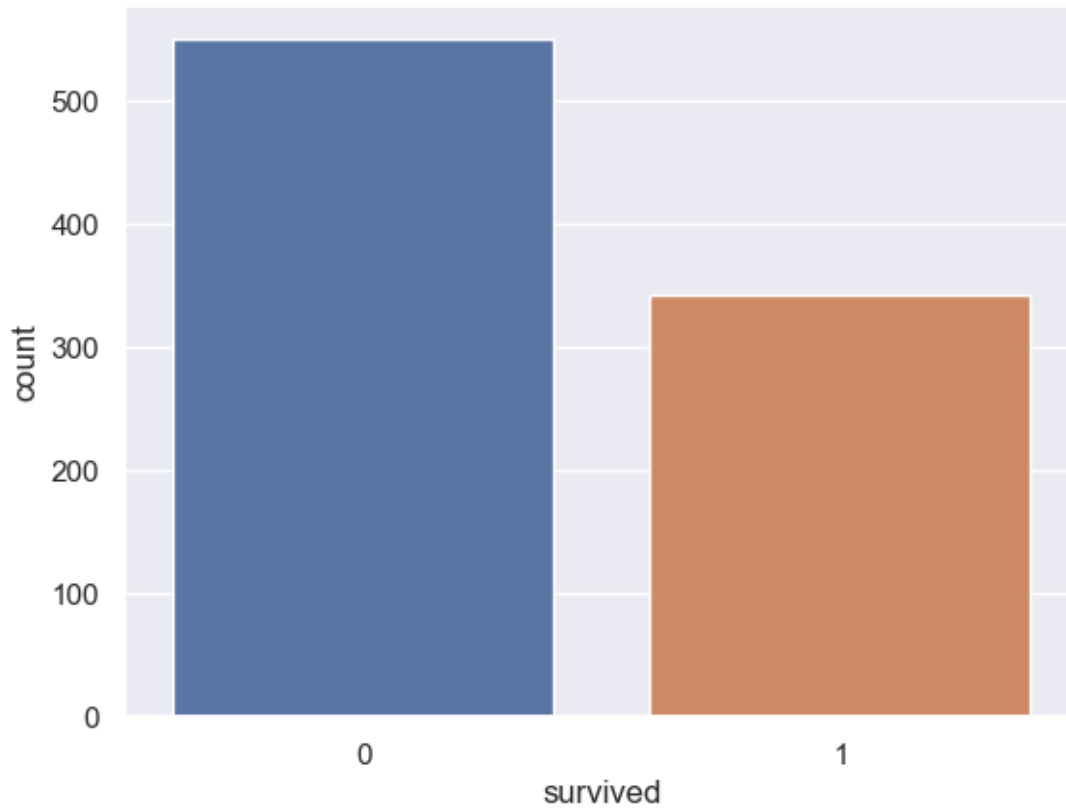
```
[11]: (891, 15)
```

Making a Count Plot

```
[12]: sns.countplot(x='class', data = titanic);
```

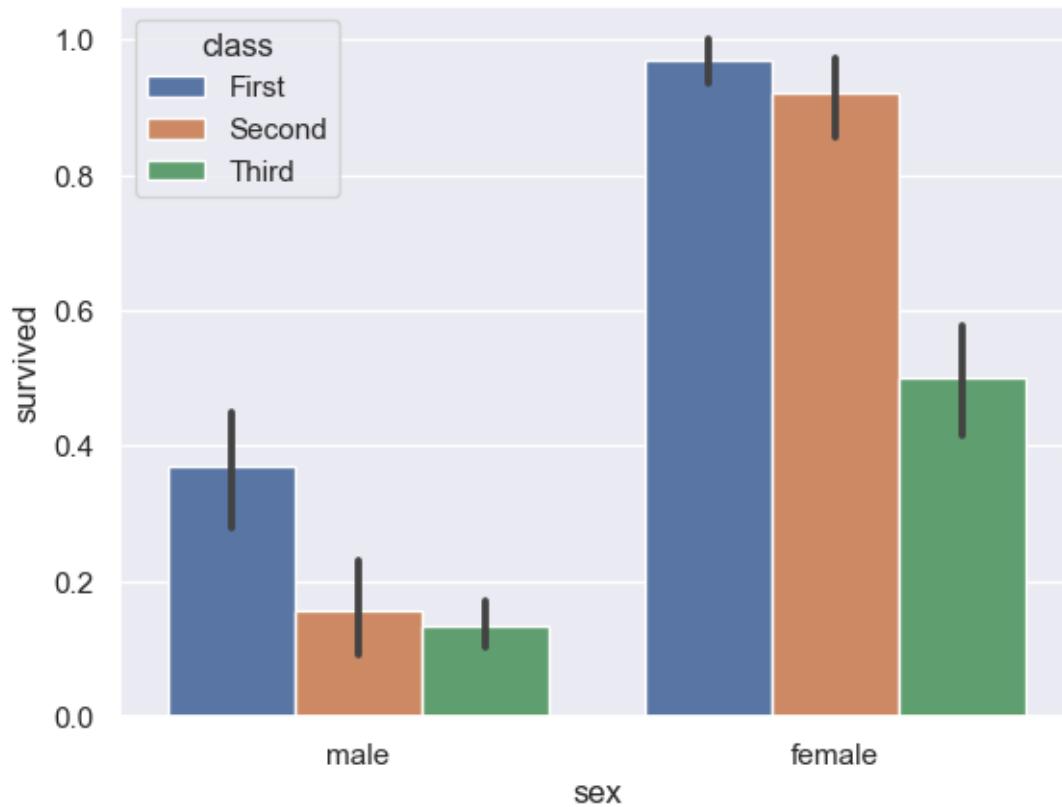


```
[13]: sns.countplot(x='survived', data = titanic);
```



Bar Chart

```
[14]: sns.barplot(x='sex', y='survived', hue = 'class', data=titanic);
```



Using house price dataset

```
[15]: boston = pd.read_csv('boston_house_prices.csv', header = 1)
      boston.head()
```

```
[15]:
```

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	\
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296	15.3	
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	

	B	LSTAT	MEDV
0	396.90	4.98	24.0
1	396.90	9.14	21.6
2	392.83	4.03	34.7
3	394.63	2.94	33.4
4	396.90	5.33	36.2

```
[16]: # Renaming a column in-place
      boston.rename(columns={'MEDV': 'PRICE'}, inplace=True)
```

```
[25]: boston.head()
```

```
[25]:
```

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	\
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296	15.3	
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	

	B	LSTAT	PRICE
0	396.90	4.98	24.0
1	396.90	9.14	21.6
2	392.83	4.03	34.7
3	394.63	2.94	33.4
4	396.90	5.33	36.2

Looking at the distribution plot

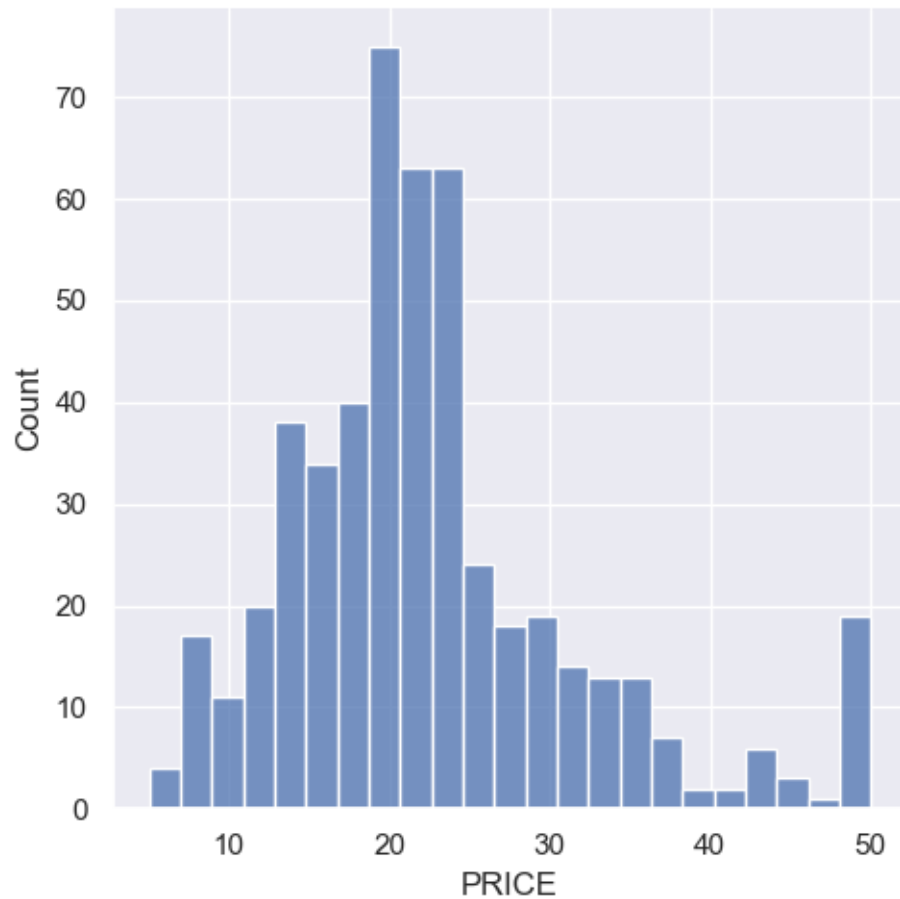
```
[18]: sns.displot(boston['PRICE']);
```

C:\Users\DONATUS\anaconda24\Lib\site-packages\seaborn\axisgrid.py:118:

UserWarning: The figure layout has changed to tight

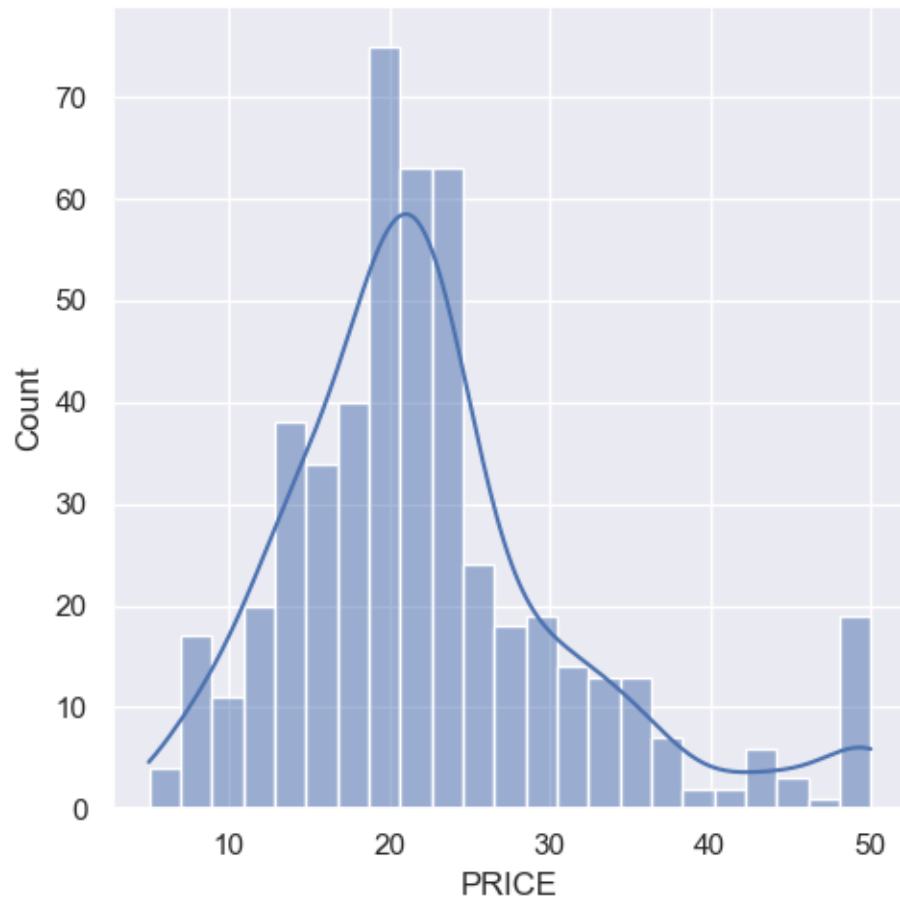
```
self._figure.tight_layout(*args, **kwargs)
```





```
[24]: sns.displot(boston['PRICE'], kde=True);
```

```
C:\Users\DONATUS\anaconda24\Lib\site-packages\seaborn\axisgrid.py:118:  
UserWarning: The figure layout has changed to tight  
self._figure.tight_layout(*args, **kwargs)
```



CORRELATION -Positive -Negative

Using heat map to plot the correlation

```
[28]: correlation = boston.corr()
      correlation
```

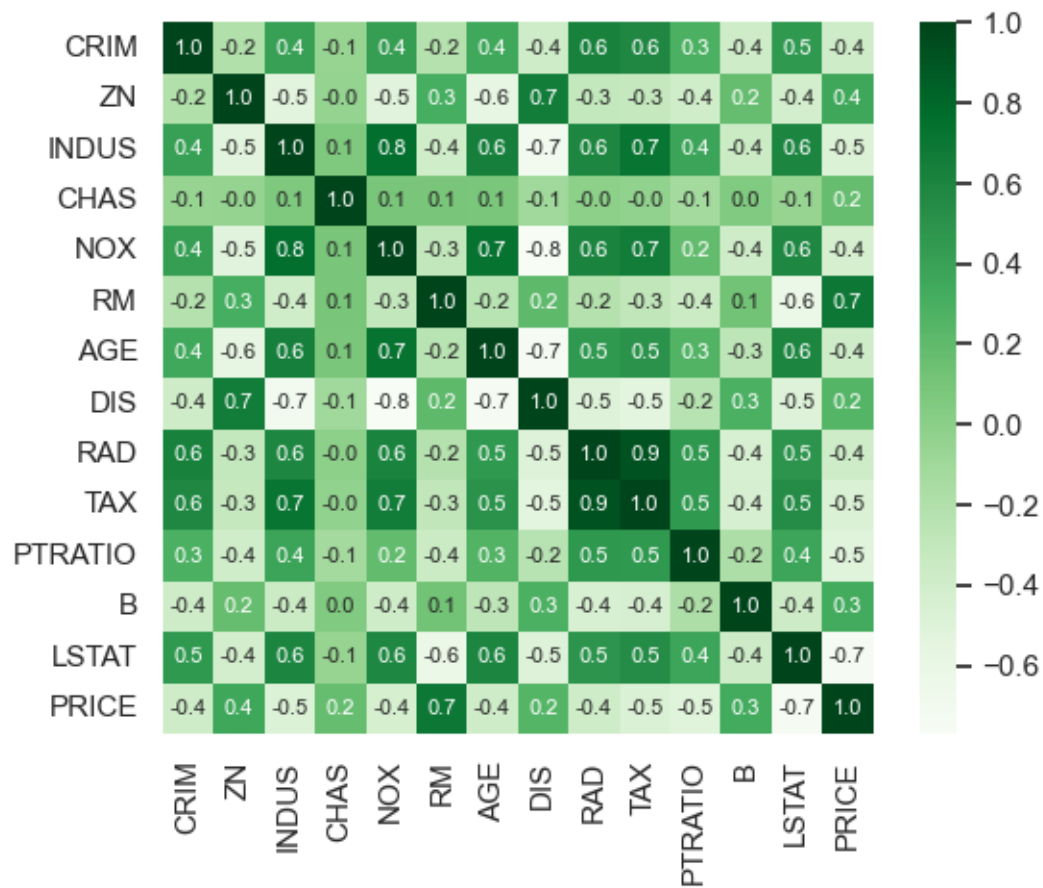
```
[28]:
```

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	\
CRIM	1.000000	-0.200469	0.406583	-0.055892	0.420972	-0.219247	0.352734	
ZN	-0.200469	1.000000	-0.533828	-0.042697	-0.516604	0.311991	-0.569537	
INDUS	0.406583	-0.533828	1.000000	0.062938	0.763651	-0.391676	0.644779	
CHAS	-0.055892	-0.042697	0.062938	1.000000	0.091203	0.091251	0.086518	
NOX	0.420972	-0.516604	0.763651	0.091203	1.000000	-0.302188	0.731470	
RM	-0.219247	0.311991	-0.391676	0.091251	-0.302188	1.000000	-0.240265	
AGE	0.352734	-0.569537	0.644779	0.086518	0.731470	-0.240265	1.000000	
DIS	-0.379670	0.664408	-0.708027	-0.099176	-0.769230	0.205246	-0.747881	
RAD	0.625505	-0.311948	0.595129	-0.007368	0.611441	-0.209847	0.456022	
TAX	0.582764	-0.314563	0.720760	-0.035587	0.668023	-0.292048	0.506456	
PTRATIO	0.289946	-0.391679	0.383248	-0.121515	0.188933	-0.355501	0.261515	

B	-0.385064	0.175520	-0.356977	0.048788	-0.380051	0.128069	-0.273534
LSTAT	0.455621	-0.412995	0.603800	-0.053929	0.590879	-0.613808	0.602339
PRICE	-0.388305	0.360445	-0.483725	0.175260	-0.427321	0.695360	-0.376955

	DIS	RAD	TAX	PTRATIO	B	LSTAT	PRICE
CRIM	-0.379670	0.625505	0.582764	0.289946	-0.385064	0.455621	-0.388305
ZN	0.664408	-0.311948	-0.314563	-0.391679	0.175520	-0.412995	0.360445
INDUS	-0.708027	0.595129	0.720760	0.383248	-0.356977	0.603800	-0.483725
CHAS	-0.099176	-0.007368	-0.035587	-0.121515	0.048788	-0.053929	0.175260
NOX	-0.769230	0.611441	0.668023	0.188933	-0.380051	0.590879	-0.427321
RM	0.205246	-0.209847	-0.292048	-0.355501	0.128069	-0.613808	0.695360
AGE	-0.747881	0.456022	0.506456	0.261515	-0.273534	0.602339	-0.376955
DIS	1.000000	-0.494588	-0.534432	-0.232471	0.291512	-0.496996	0.249929
RAD	-0.494588	1.000000	0.910228	0.464741	-0.444413	0.488676	-0.381626
TAX	-0.534432	0.910228	1.000000	0.460853	-0.441808	0.543993	-0.468536
PTRATIO	-0.232471	0.464741	0.460853	1.000000	-0.177383	0.374044	-0.507787
B	0.291512	-0.444413	-0.441808	-0.177383	1.000000	-0.366087	0.333461
LSTAT	-0.496996	0.488676	0.543993	0.374044	-0.366087	1.000000	-0.737663
PRICE	0.249929	-0.381626	-0.468536	-0.507787	0.333461	-0.737663	1.000000

```
[49]: sns.heatmap(correlation,
                  cbar=True, # Show the colorbar
                  square=True, # Make the heatmap square
                  fmt='.1f', # Format the values in the heatmap
                  annot=True, # Add annotations to the heatmap
                  annot_kws={'size': 8}, # Set the annotation font size
                  cmap='Greens'); # Set the colormap to Blues
```



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