

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

#simple imputer
array = np.array([[1,2],[np.nan,3],[7,6]])
from sklearn.impute import SimpleImputer
imputer = SimpleImputer(missing_values=np.nan, strategy="mean")
imputed_array = imputer.fit_transform(array)
print(imputed_array)
```

```
[[1. 2.]
 [4. 3.]
 [7. 6.]]
```

```
#minmaxscaler
from sklearn.preprocessing import MinMaxScaler
array_2 = np.array([[1,2],[6,7],[3,4]])
scaler = MinMaxScaler()
scaled_array = scaler.fit_transform(array_2)
print(scaled_array)
```

```
[[0.  0. ]
 [1.  1. ]
 [0.4 0.4]]
```

```
data = pd.read_csv("iris.csv")
df = pd.DataFrame(data)
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  -
0   sepallength     150 non-null   float64
1   sepalwidth      150 non-null   float64
2   petallength     150 non-null   float64
3   petalwidth      150 non-null   float64
4   class           150 non-null   object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
df_nonull = df[df.isnull()==False]
df_nonull
```

	sepallength	sepalwidth	petallength	petalwidth	class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa

4	5.0	3.6	1.4	0.2	Iris-setosa
...
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

[150 rows x 5 columns]

df_nonull.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 150 entries, 0 to 149

Data columns (total 5 columns):

#	Column	Non-Null Count	Dtype
0	sepalwidth	150 non-null	float64
1	sepalwidth	150 non-null	float64
2	petallength	150 non-null	float64
3	petalwidth	150 non-null	float64
4	class	150 non-null	object

dtypes: float64(4), object(1)

memory usage: 6.0+ KB

df.head(5)

	sepalwidth	sepalwidth	petallength	petalwidth	class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

df.tail(5)

	sepalwidth	sepalwidth	petallength	petalwidth	class
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

df["petallength"].dtype

dtype('float64')

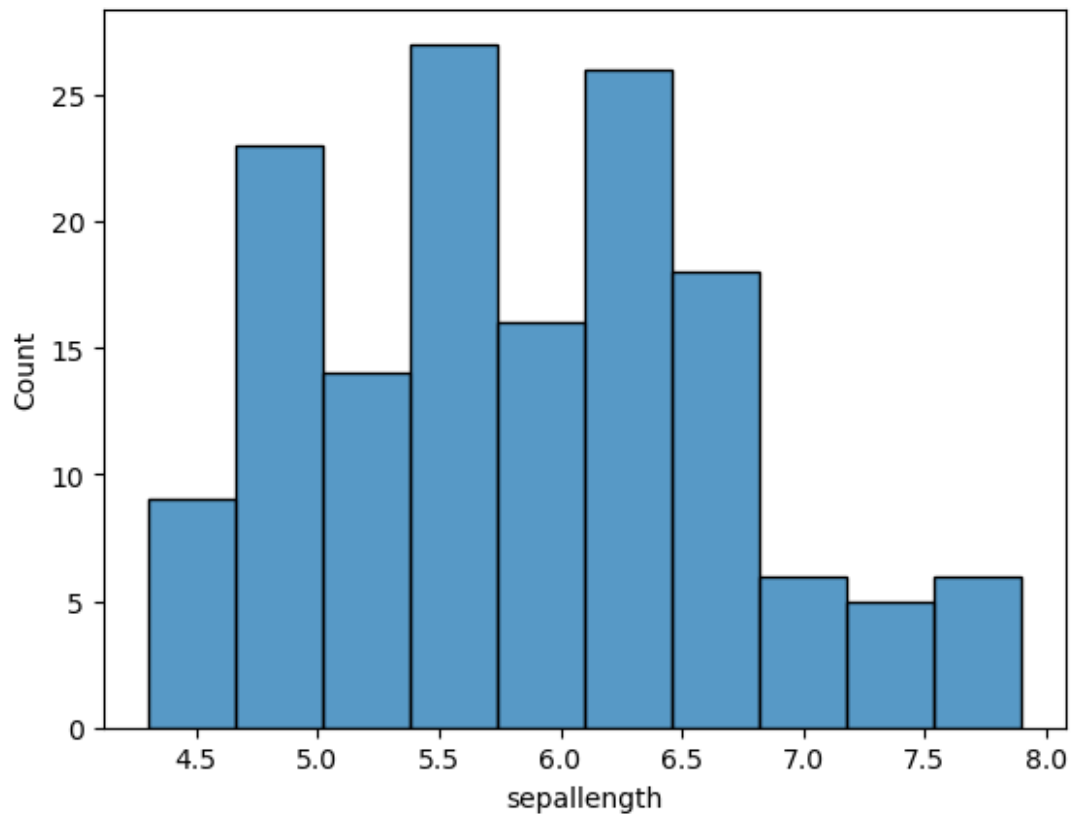
df["sepalwidth"].mean()

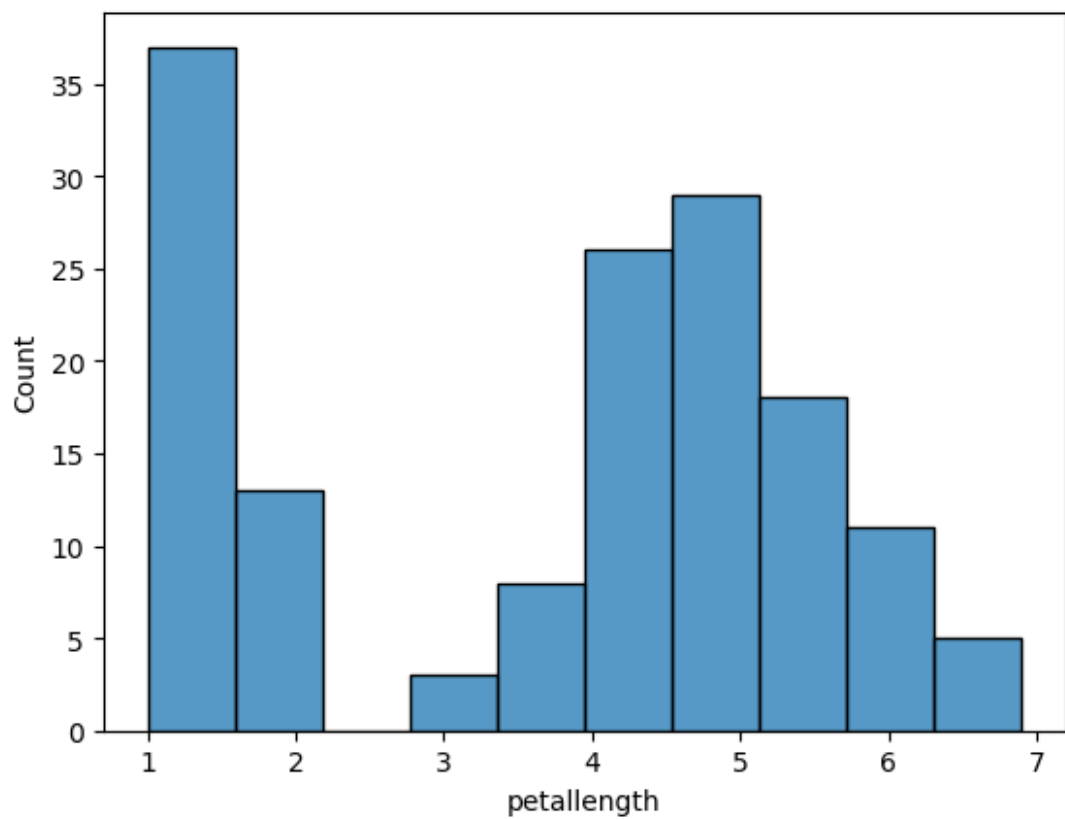
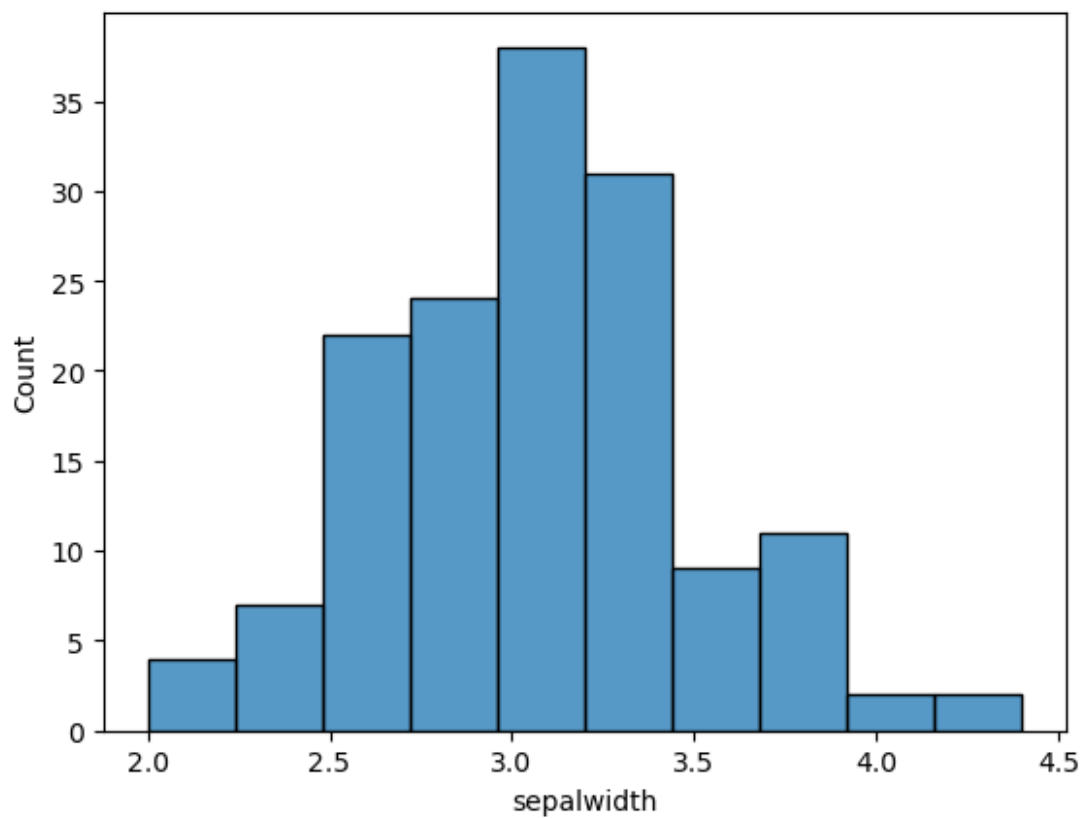
5.843333333333334

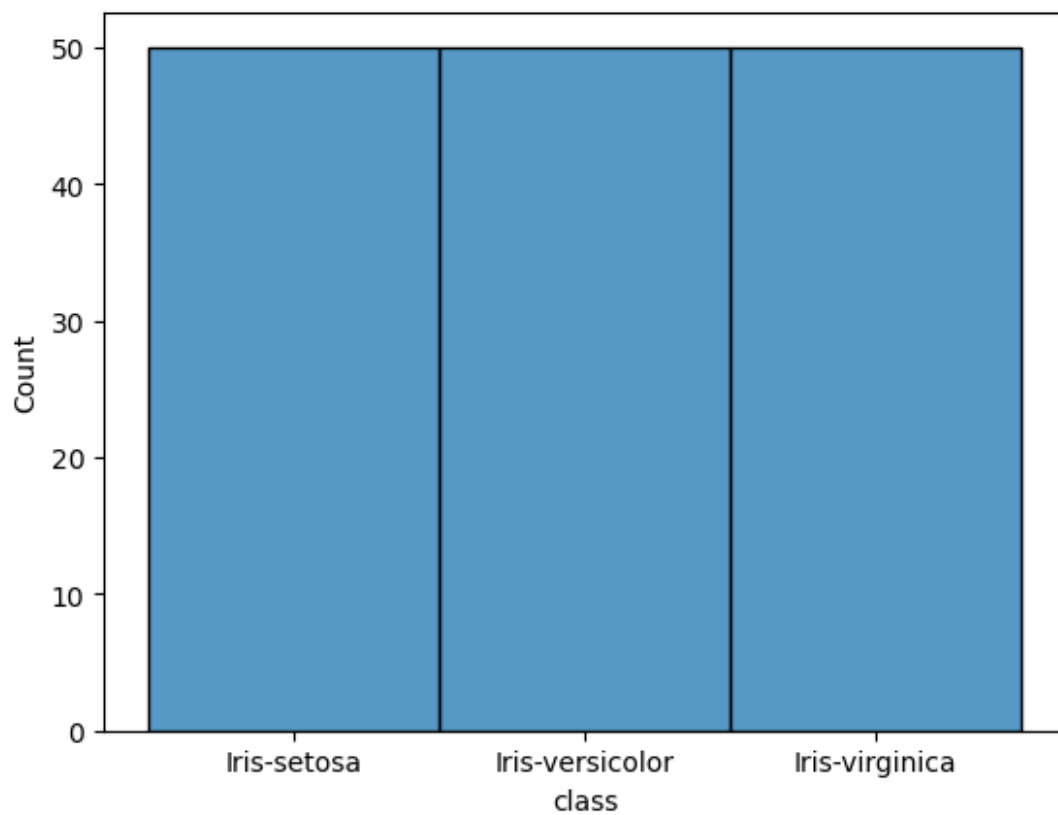
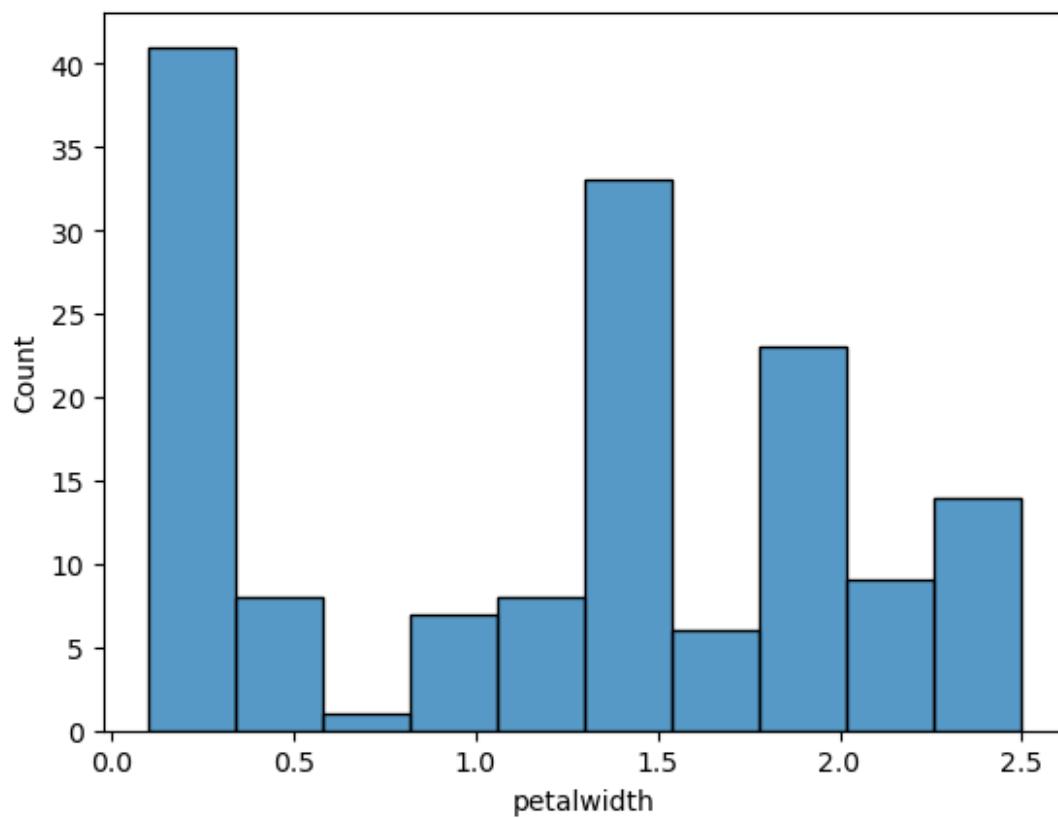
df["petalwidth"].std()

0.7631607417008414

```
import seaborn as sns  
for i in list(df.columns):  
    sns.histplot(df[i],bins=10)  
    plt.show()
```

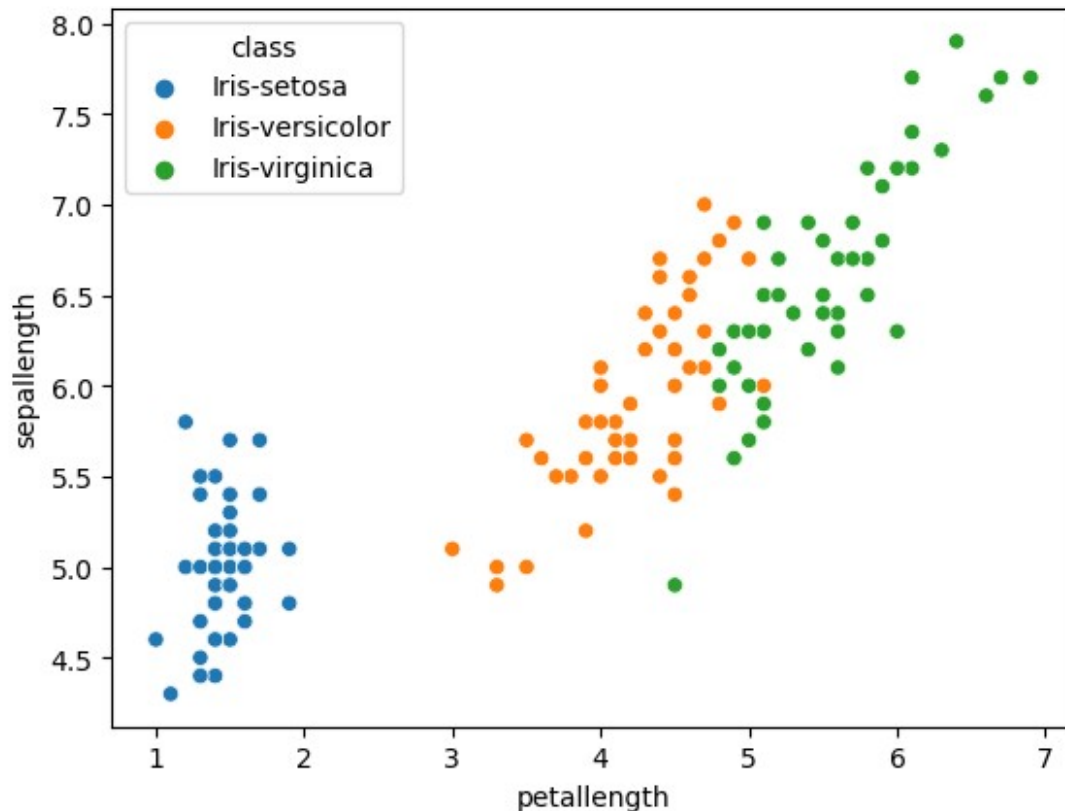






```
sns.scatterplot(data=df,x=df["petallength"],y=df["sepalength"],hue=df["class"])
```

```
<Axes: xlabel='petallength', ylabel='sepalength'>
```



```
wine = pd.read_csv("WineQT.csv")
df_wine = pd.DataFrame(wine)
df_wine.head()
```

	fixed_acidity	volatile_acidity	citric_acid	residual_sugar
0	7.4	0.70	0.00	1.9
1	7.8	0.88	0.00	2.6
2	7.8	0.76	0.04	2.3
3	11.2	0.28	0.56	1.9
4	7.4	0.70	0.00	1.9

	free_sulfur_dioxide	total_sulfur_dioxide	density	pH	sulphates
0	171	2150	0.9948	3.51	0.09
1	178	1610	0.9981	3.26	0.16
2	179	2500	0.9968	3.17	0.19
3	244	3690	0.9999	4.01	0.27
4	159	1720	0.9956	3.42	0.09

0	11.0	34.0	0.9978	3.51	0.56
1	25.0	67.0	0.9968	3.20	0.68
2	15.0	54.0	0.9970	3.26	0.65
3	17.0	60.0	0.9980	3.16	0.58
4	11.0	34.0	0.9978	3.51	0.56

	alcohol	quality	Id
0	9.4	5	0
1	9.8	5	1
2	9.8	5	2
3	9.8	6	3
4	9.4	5	4

```
df_wine.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 1143 entries, 0 to 1142
```

```
Data columns (total 13 columns):
```

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	fixed_acidity	1143 non-null	float64
1	volatile_acidity	1143 non-null	float64
2	citric_acid	1143 non-null	float64
3	residual_sugar	1143 non-null	float64
4	chlorides	1143 non-null	float64
5	free_sulfur_dioxide	1143 non-null	float64
6	total_sulfur_dioxide	1143 non-null	float64
7	density	1143 non-null	float64
8	pH	1143 non-null	float64
9	sulphates	1143 non-null	float64
10	alcohol	1143 non-null	float64
11	quality	1143 non-null	int64
12	Id	1143 non-null	int64

```
dtypes: float64(11), int64(2)
```

```
memory usage: 116.2 KB
```

```
df_wine.shape
```

```
(1143, 13)
```

```
df_wine.isnull().sum()
```

fixed_acidity	0
volatile_acidity	0
citric_acid	0
residual_sugar	0

```
chlorides          0
free_sulfur_dioxide 0
total_sulfur_dioxide 0
density            0
pH                 0
sulphates          0
alcohol            0
quality            0
Id                 0
dtype: int64
```

```
for j in list(df_wine.columns):
    print("1st quartile:",np.percentile(df_wine[j],25))
    print("2nd quartile:",np.percentile(df_wine[j],75))
    print()
```

```
1st quartile: 7.1
2nd quartile: 9.1
```

```
1st quartile: 0.3925
2nd quartile: 0.64
```

```
1st quartile: 0.09
2nd quartile: 0.42
```

```
1st quartile: 1.9
2nd quartile: 2.6
```

```
1st quartile: 0.07
2nd quartile: 0.09
```

```
1st quartile: 7.0
2nd quartile: 21.0
```

```
1st quartile: 21.0
2nd quartile: 61.0
```

```
1st quartile: 0.99557
2nd quartile: 0.997845
```

```
1st quartile: 3.205
2nd quartile: 3.4
```

```
1st quartile: 0.55
2nd quartile: 0.73
```

```
1st quartile: 9.5
2nd quartile: 11.1
```

```
1st quartile: 5.0
2nd quartile: 6.0
```

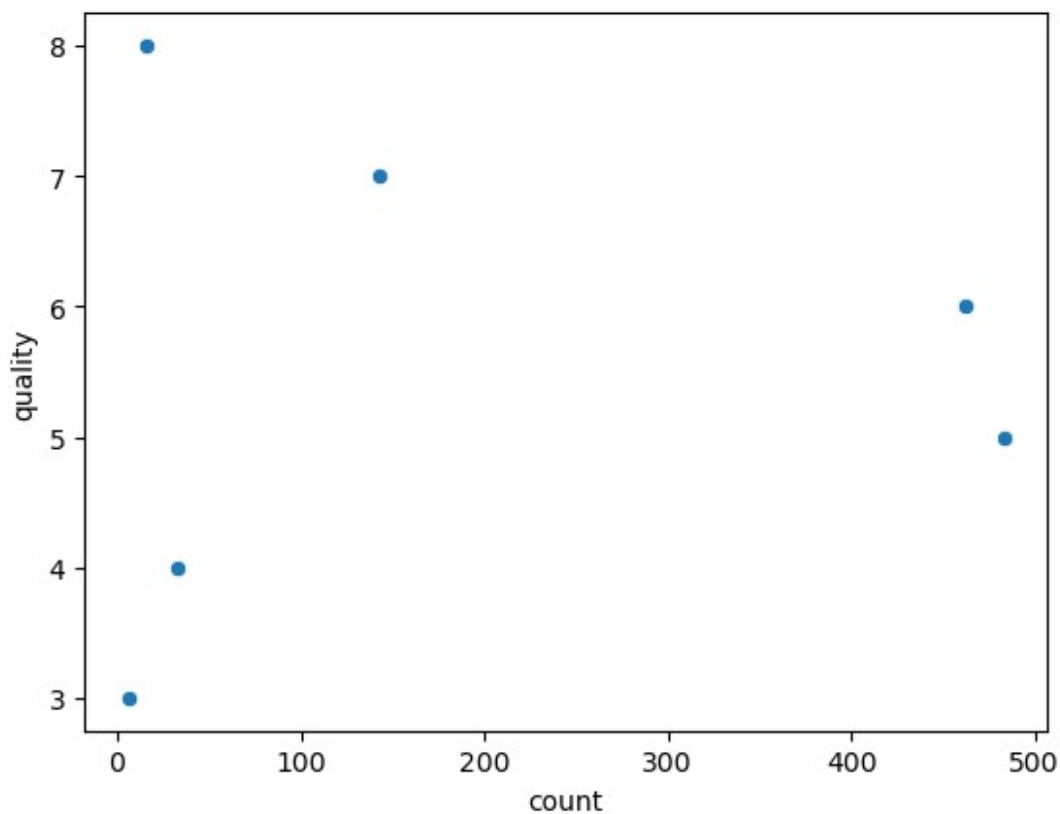


```
1st quartile: 411.0  
2nd quartile: 1209.5
```

```
df_plot = df_wine["quality"].value_counts().sort_index().reset_index()  
df_plot.columns = ["quality", "count"]  
df_plot
```

	quality	count
0	3	6
1	4	33
2	5	483
3	6	462
4	7	143
5	8	16

```
sns.scatterplot(data =  
df_plot,x=df_plot["count"],y=df_plot["quality"])  
plt.show()
```



```
sns.barplot(data=df_wine,x=df_wine["quality"],y=df_wine["volatile_acid  
ity"])
```

```
<Axes: xlabel='quality', ylabel='volatile_acidity'>
```

