

```
In [1]: # Importing packages
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
#To ignore warnings
import warnings
warnings.filterwarnings('ignore')
```

```
In [3]: iris = pd.read_csv(r"C:\Users\user\Downloads\iris.data",header=None)
iris
```

```
Out[3]:
```

	0	1	2	3	4
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 × 5 columns

```
In [5]: iris.columns = ['SL','SW','PL','PW','Flower']
iris.head()
```

```
Out[5]:
```

	SL	SW	PL	PW	Flower
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

```
In [7]: iris.isnull().sum()
```

```
Out[7]: SL      0
        SW      0
        PL      0
        PW      0
        Flower   0
        dtype: int64
```

```
In [11]: iris.dtypes
```

```
Out[11]: SL      float64
        SW      float64
        PL      float64
        PW      float64
        Flower   object
        dtype: object
```

```
In [13]: for i in iris.columns:
        print(i,':','\n',iris[i].unique(),'\n')
```

```
SL :
[5.1 4.9 4.7 4.6 5.  5.4 4.4 4.8 4.3 5.8 5.7 5.2 5.5 4.5 5.3 7.  6.4 6.9
 6.5 6.3 6.6 5.9 6.  6.1 5.6 6.7 6.2 6.8 7.1 7.6 7.3 7.2 7.7 7.4 7.9]
```

```
SW :
[3.5 3.  3.2 3.1 3.6 3.9 3.4 2.9 3.7 4.  4.4 3.8 3.3 4.1 4.2 2.3 2.8 2.4
 2.7 2.  2.2 2.5 2.6]
```

```
PL :
[1.4 1.3 1.5 1.7 1.6 1.1 1.2 1.  1.9 4.7 4.5 4.9 4.  4.6 3.3 3.9 3.5 4.2
 3.6 4.4 4.1 4.8 4.3 5.  3.8 3.7 5.1 3.  6.  5.9 5.6 5.8 6.6 6.3 6.1 5.3
 5.5 6.7 6.9 5.7 6.4 5.4 5.2]
```

```
PW :
[0.2 0.4 0.3 0.1 0.5 0.6 1.4 1.5 1.3 1.6 1.  1.1 1.8 1.2 1.7 2.5 1.9 2.1
 2.2 2.  2.4 2.3]
```

```
Flower :
['Iris-setosa' 'Iris-versicolor' 'Iris-virginica']
```

```
In [17]: iris.shape
```

```
Out[17]: (150, 5)
```

```
In [15]: iris.describe()
```

Out[15]:

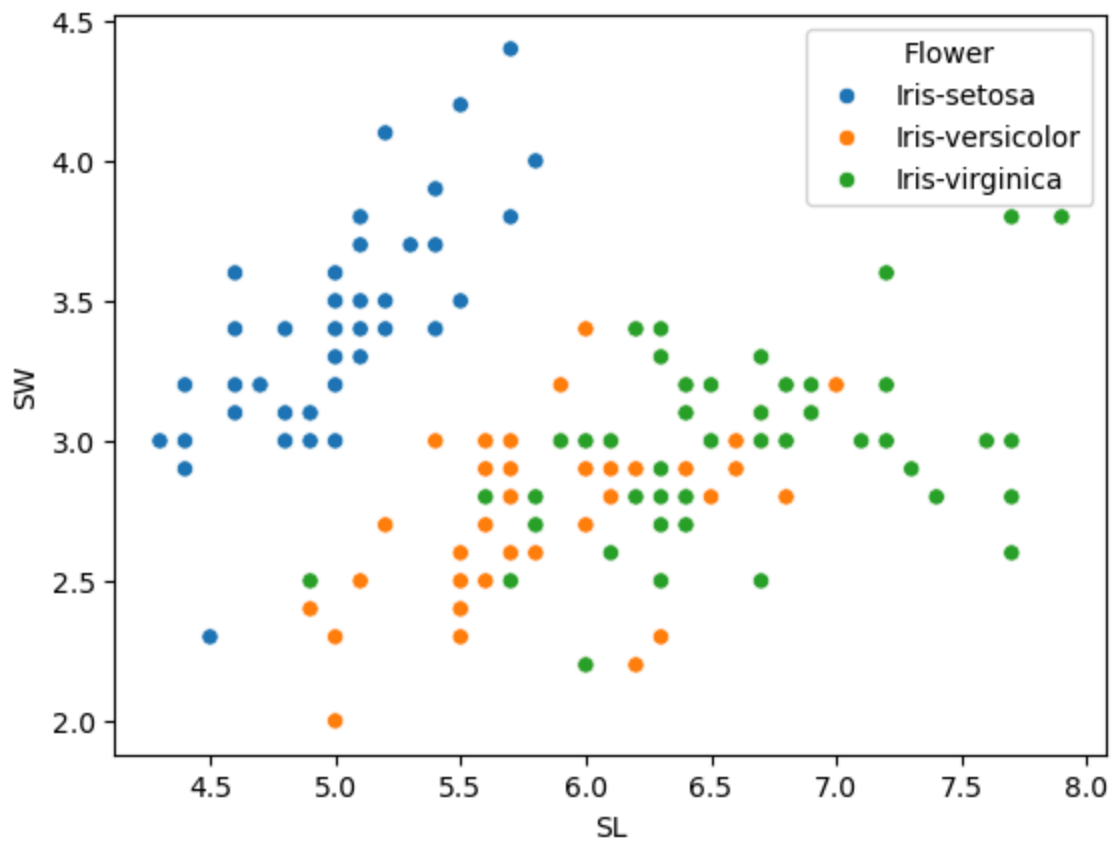
	SL	SW	PL	PW
<b>count</b>	150.000000	150.000000	150.000000	150.000000
<b>mean</b>	5.843333	3.054000	3.758667	1.198667
<b>std</b>	0.828066	0.433594	1.764420	0.763161
<b>min</b>	4.300000	2.000000	1.000000	0.100000
<b>25%</b>	5.100000	2.800000	1.600000	0.300000
<b>50%</b>	5.800000	3.000000	4.350000	1.300000
<b>75%</b>	6.400000	3.300000	5.100000	1.800000
<b>max</b>	7.900000	4.400000	6.900000	2.500000

```
In [21]: iris.groupby('Flower').mean()
```

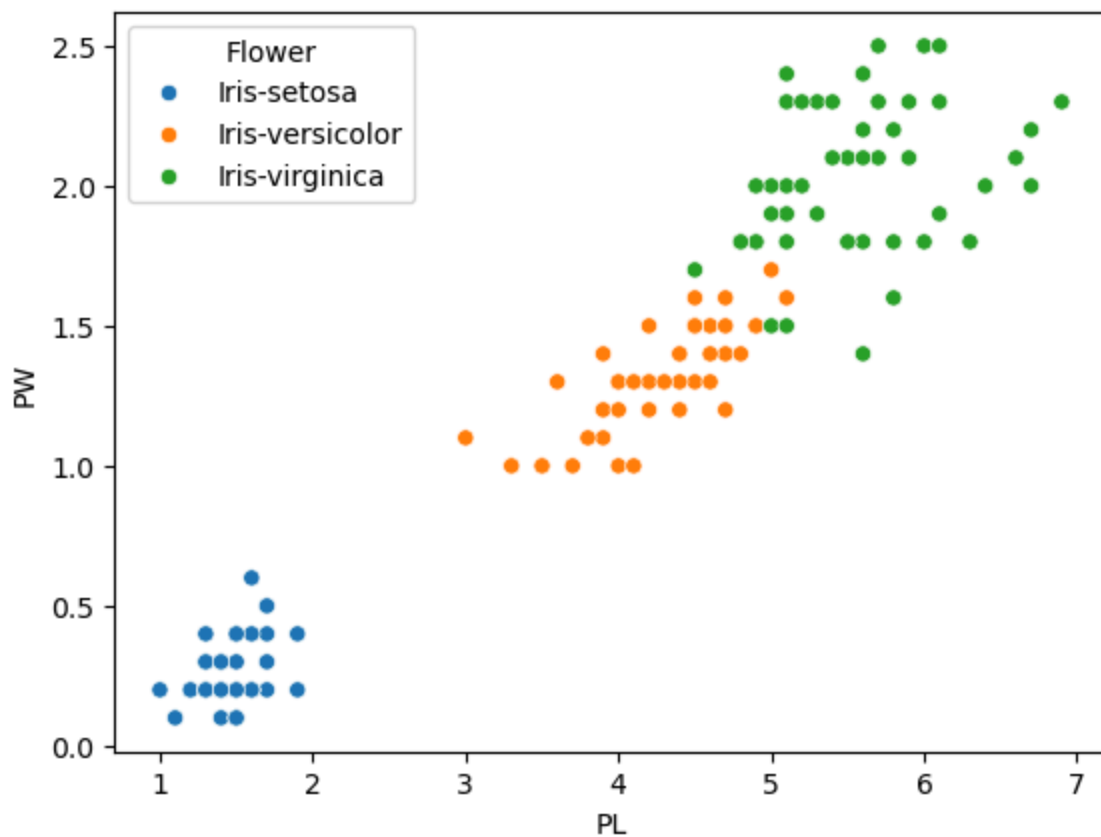
Out[21]:

	SL	SW	PL	PW
<b>Flower</b>				
<b>Iris-setosa</b>	5.006	3.418	1.464	0.244
<b>Iris-versicolor</b>	5.936	2.770	4.260	1.326
<b>Iris-virginica</b>	6.588	2.974	5.552	2.026

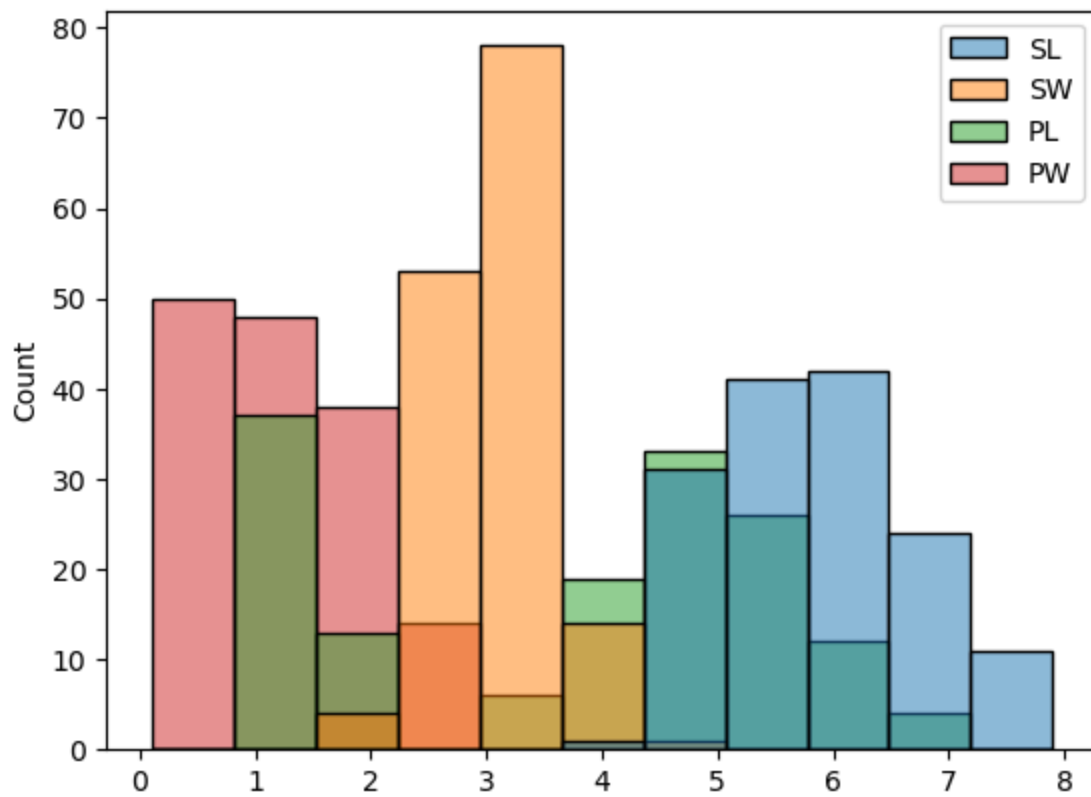
```
In [25]: sns.scatterplot(x='SL', y='SW', hue='Flower', data=iris)
plt.show()
```



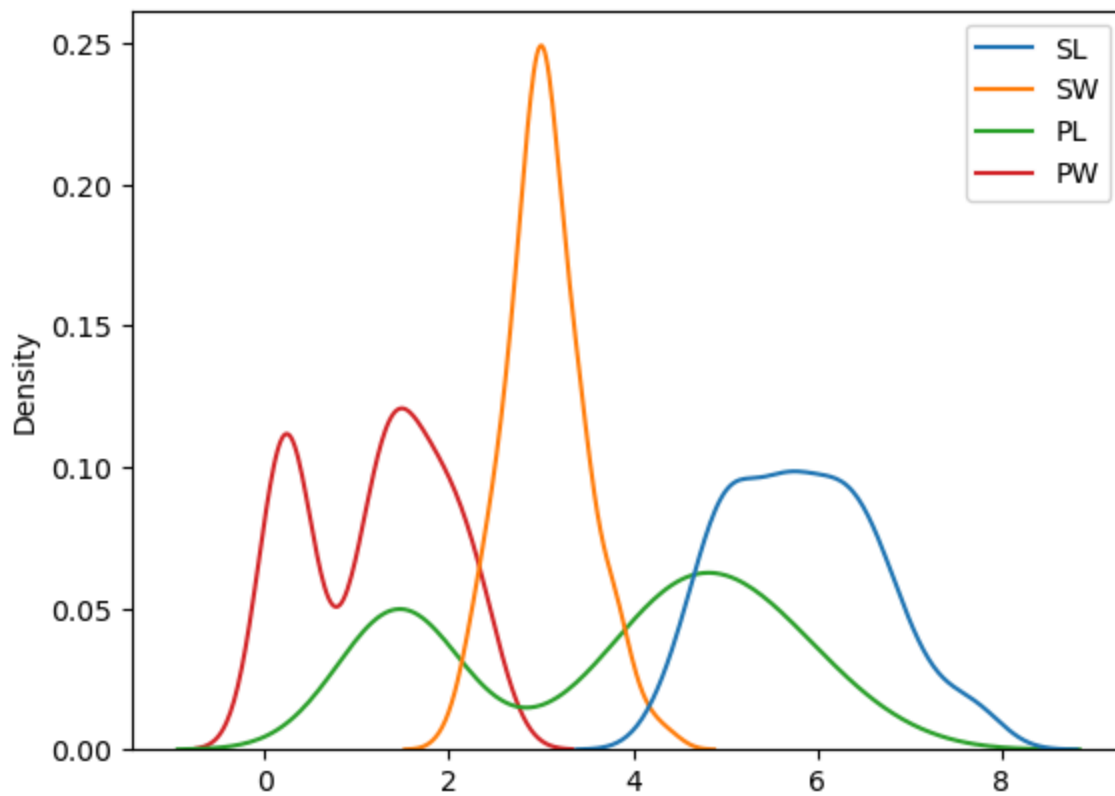
```
In [27]: sns.scatterplot(x='PL', y='PW', hue='Flower', data=iris)
plt.show()
```



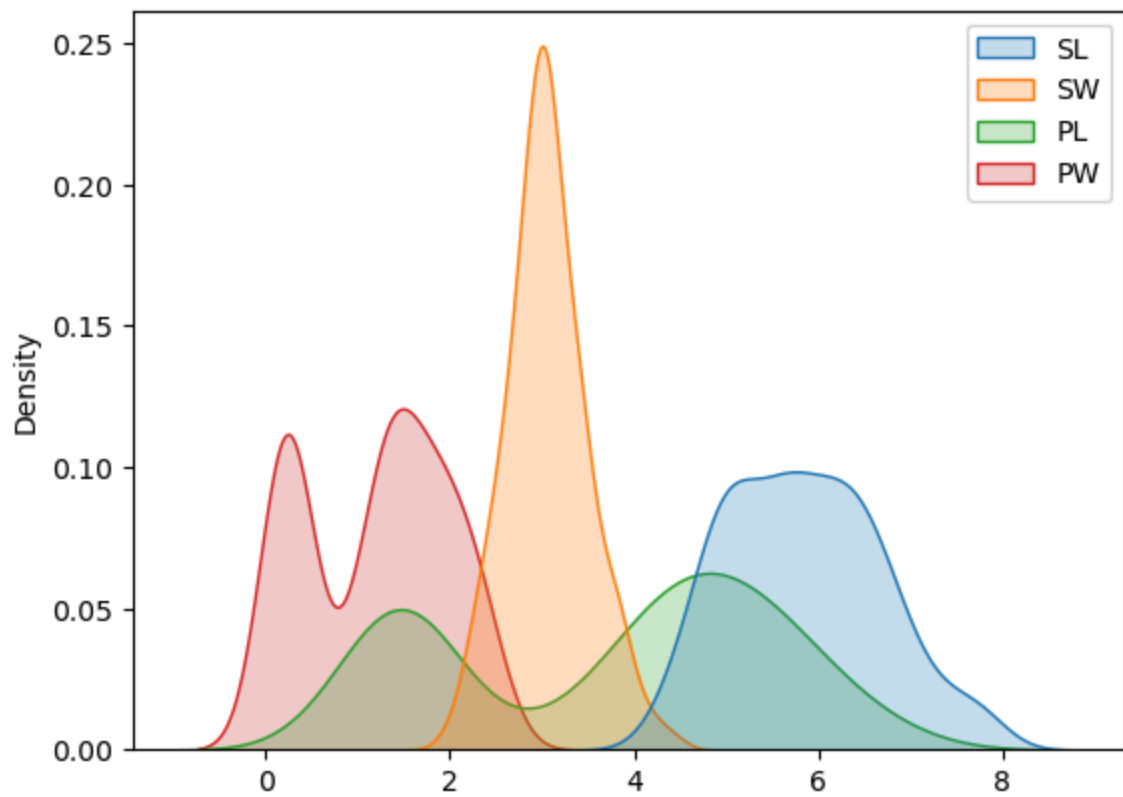
```
In [29]: sns.histplot(data=iris)
plt.show()
```



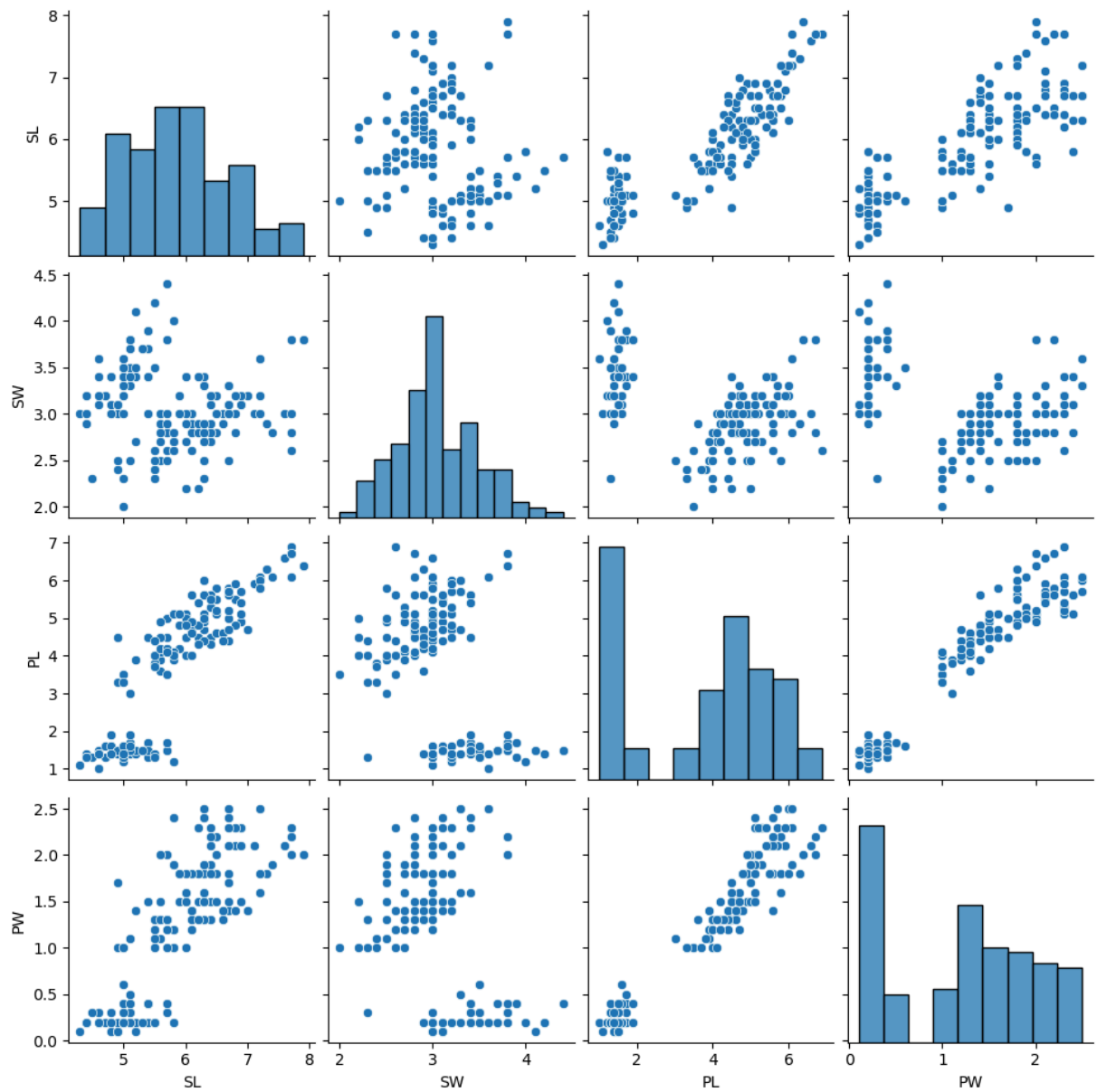
```
In [31]: sns.kdeplot(data=iris)
plt.show()
```



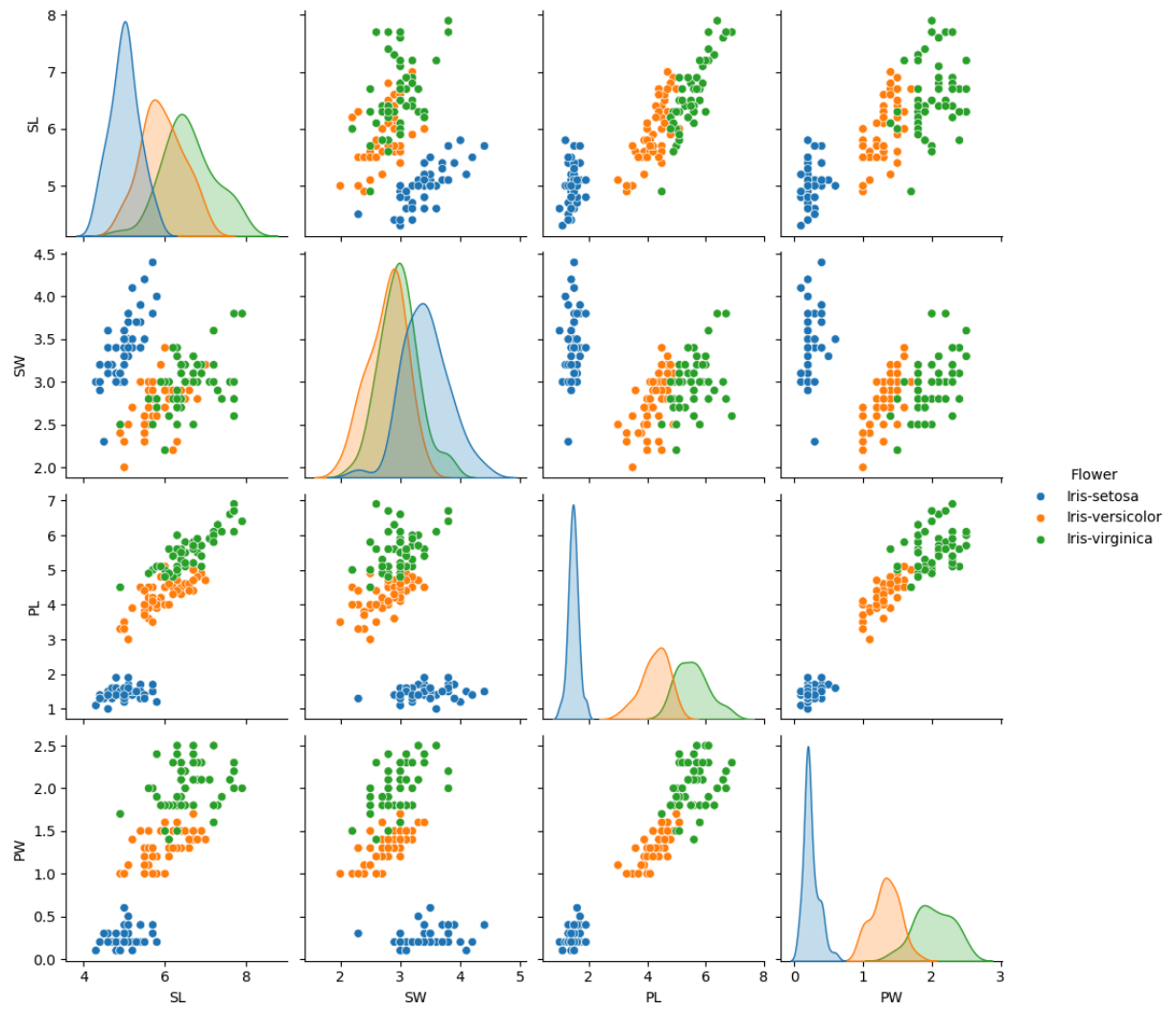
```
In [33]: sns.kdeplot(data=iris,fill=True)  
plt.show()
```



```
In [35]: sns.pairplot(iris)  
plt.show()
```

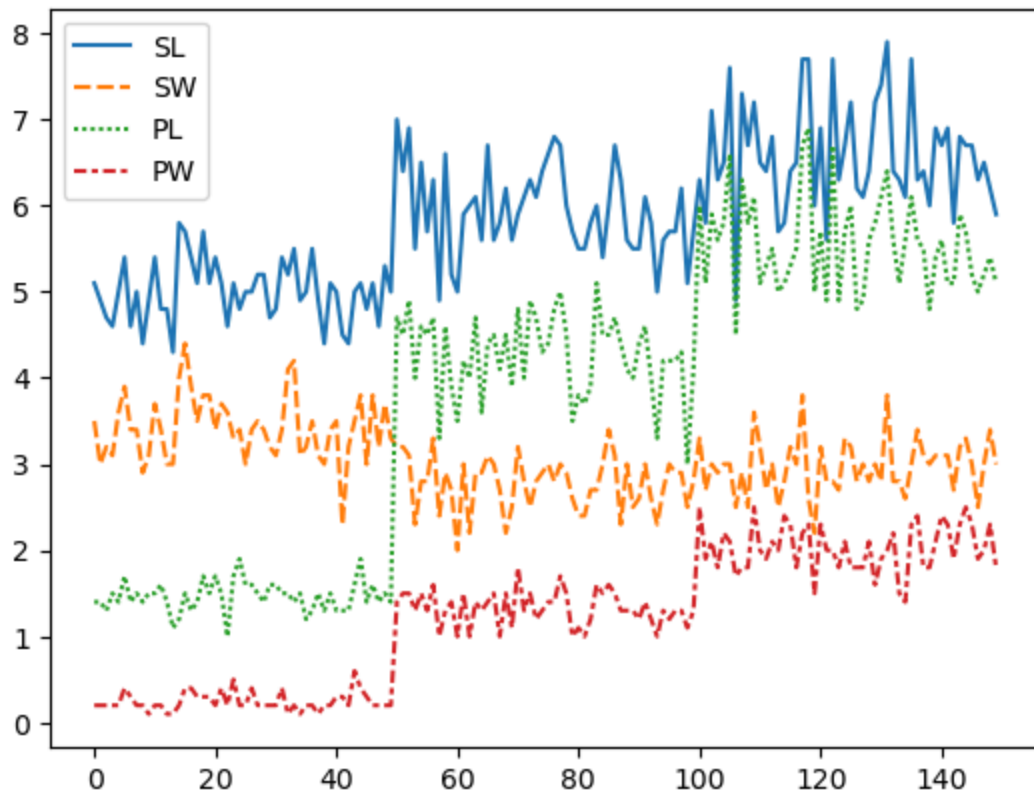


```
In [37]: sns.pairplot(iris,hue='Flower')
plt.show()
```

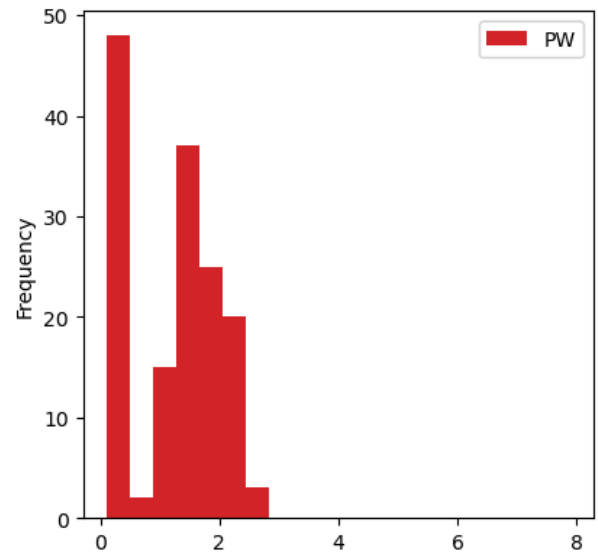
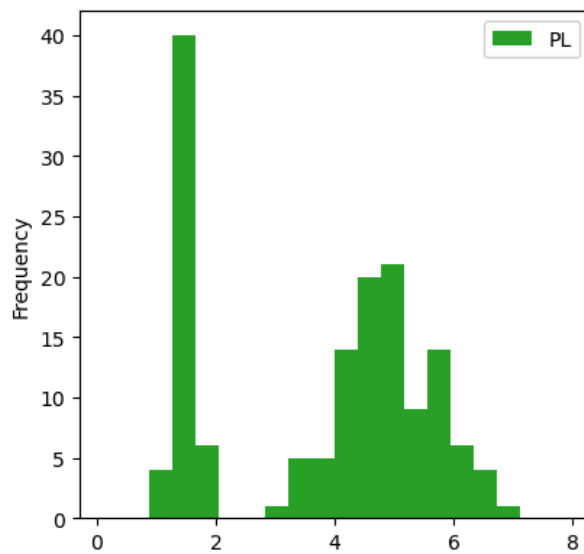
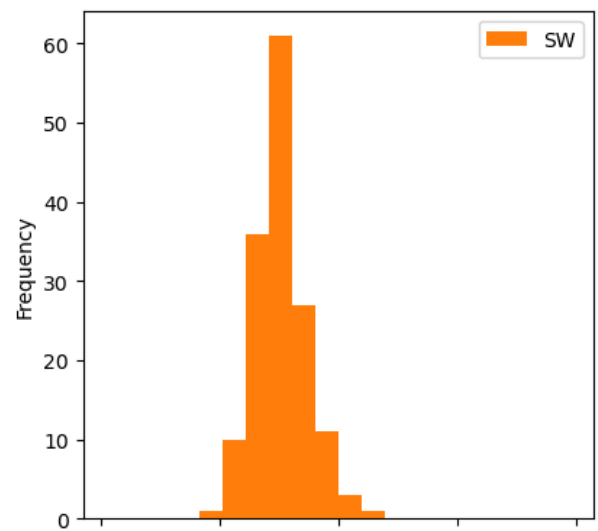
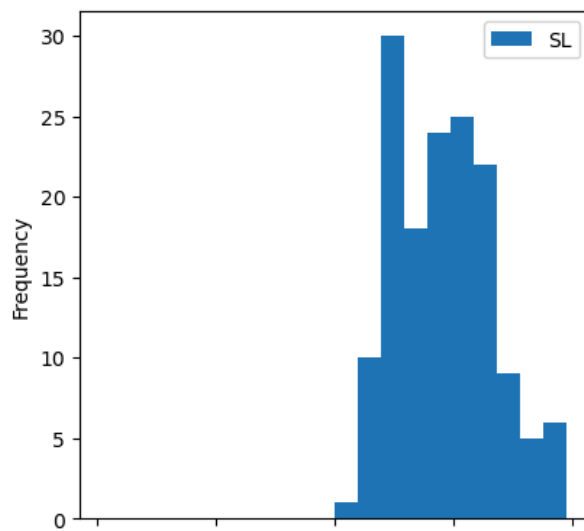


```
In [39]: sns.lineplot(data=iris.drop(['Flower'], axis=1))
plt.show()
```

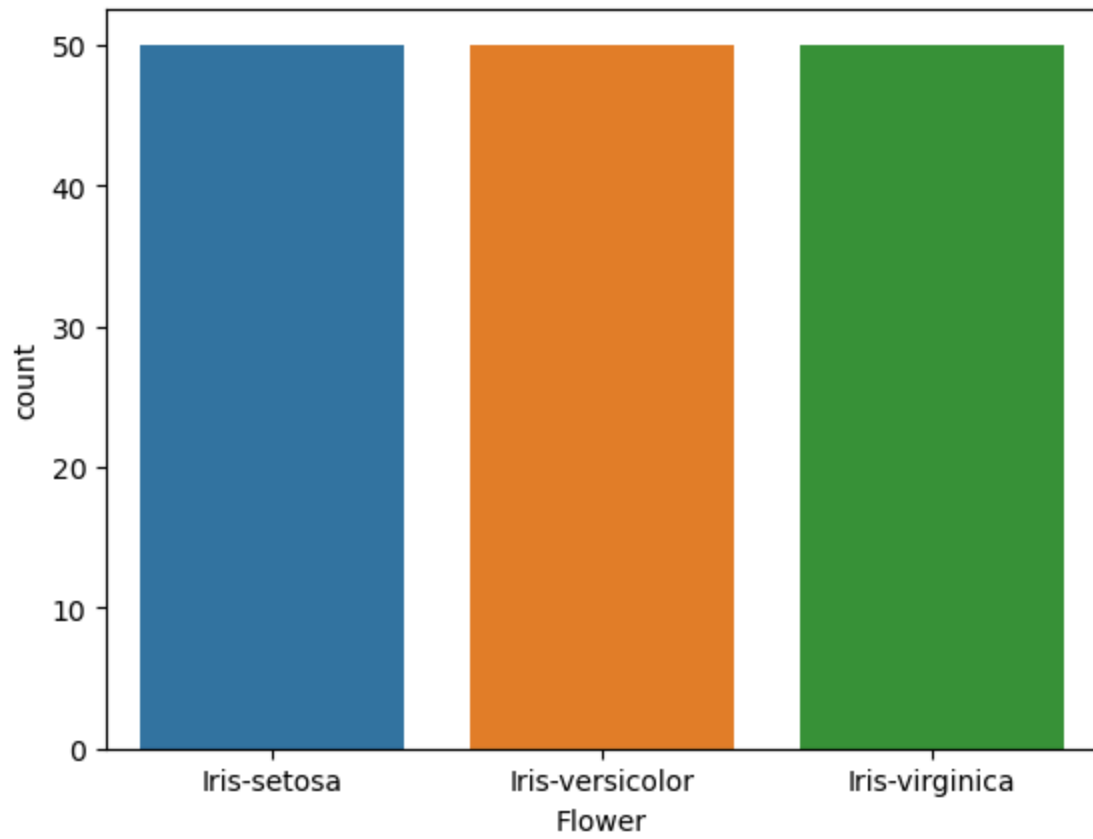




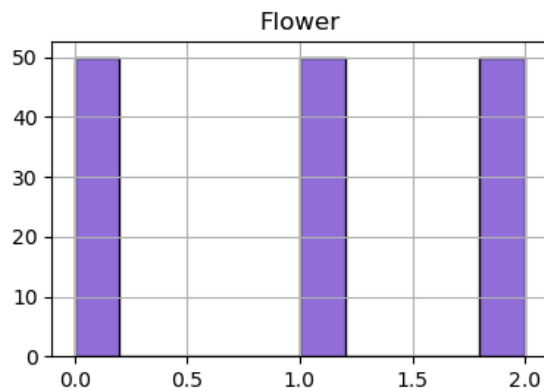
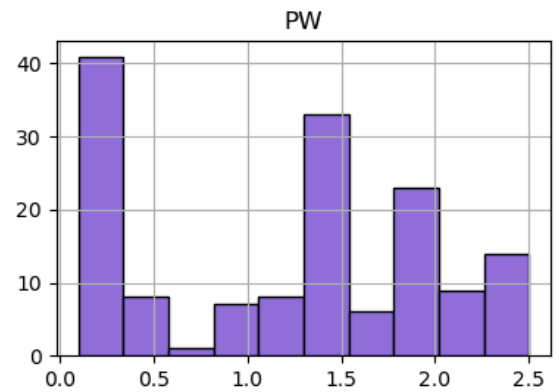
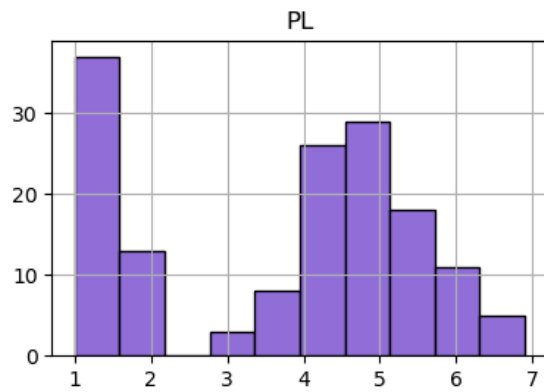
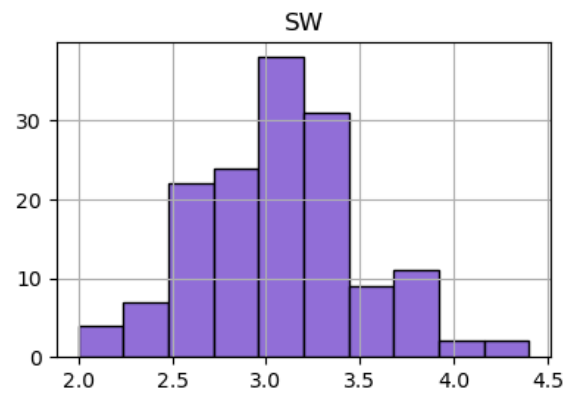
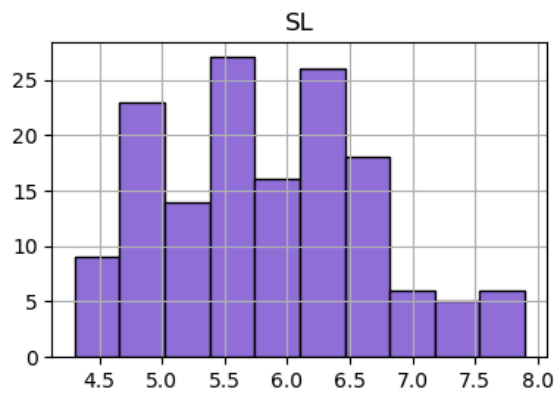
```
In [41]: iris.plot.hist(subplots=True, layout=(2,2), figsize=(10, 10), bins=20)
plt.show()
```



```
In [45]: sns.countplot(x=iris.Flower, hue=iris.Flower)
plt.show()
```



```
In [85]: iris.hist(color= 'mediumpurple' ,edgecolor='black',figsize=(10,10))  
plt.show()
```



```
In [47]: from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
```

```
In [49]: iris.Flower = le.fit_transform(iris.Flower)
iris.Flower.unique()
```

```
Out[49]: array([0, 1, 2])
```

```
In [51]: le.inverse_transform([0,1,2])
```

```
Out[51]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)
```

```
In [53]: iris.Flower = le.fit_transform(iris.Flower)
set(iris.Flower)
```

```
Out[53]: {0, 1, 2}
```

```
In [55]: le.inverse_transform([0,1,2])
```

```
Out[55]: array([0, 1, 2])
```

```
In [59]: iris.head()
```

```
Out[59]:
```

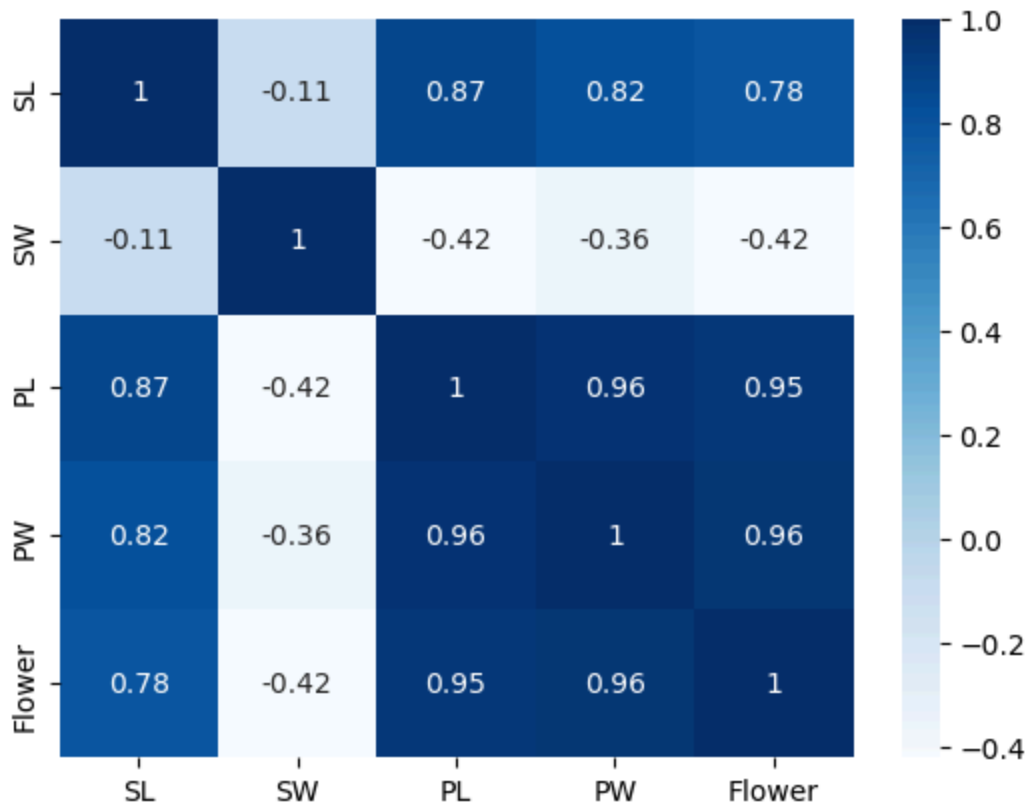
	SL	SW	PL	PW	Flower
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

```
In [61]: c = iris.corr()  
c
```

```
Out[61]:
```

	SL	SW	PL	PW	Flower
SL	1.000000	-0.109369	0.871754	0.817954	0.782561
SW	-0.109369	1.000000	-0.420516	-0.356544	-0.419446
PL	0.871754	-0.420516	1.000000	0.962757	0.949043
PW	0.817954	-0.356544	0.962757	1.000000	0.956464
Flower	0.782561	-0.419446	0.949043	0.956464	1.000000

```
In [63]: sns.heatmap(c,annot=True,cmap='Blues')  
plt.show()
```



```
In [65]: x = iris.drop('Flower',axis=1)
y = iris.Flower
```

```
In [67]: from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.3)
```

```
In [69]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()

x_train = sc.fit_transform(x_train)
x_test = sc.fit_transform(x_test)
```

```
In [71]: from sklearn.linear_model import LogisticRegression
lr = LogisticRegression()
lr.fit(x_train,y_train)
```

```
Out[71]: ▾ LogisticRegression
LogisticRegression()
```

```
In [73]: pred = lr.predict(x_test)
pred
```

```
Out[73]: array([0, 2, 1, 2, 2, 0, 0, 2, 2, 1, 2, 1, 0, 2, 2, 2, 0, 1, 2, 0, 0, 2,
                2, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 2, 2, 2, 1, 0, 0, 2, 1,
                1], dtype=int64)
```

```
In [75]: y_test
```

```
Out[75]: 37      0
        72      1
        85      1
        136     2
        134     2
        40      0
        28      0
        108     2
        139     2
        98      1
        104     2
        80      1
         0      0
        110     2
        145     2
        105     2
         3      0
        71      1
        128     2
        22      0
        39      0
        115     2
        68      1
        43      0
        24      0
        64      1
        33      0
         2      0
        45      0
        58      1
        36      0
        54      1
        30      0
        97      1
        95      1
        84      1
        77      1
        116     2
        117     2
        73      1
        21      0
        19      0
        124     2
        66      1
        88      1
        Name: Flower, dtype: int64
```

```
In [77]: from sklearn.metrics import accuracy_score, recall_score, precision_score, f1_score

ac = accuracy_score(pred, y_test)
print('Accuracy : ', ac)
re = recall_score(pred, y_test, average='weighted')
pr = precision_score(pred, y_test, average='weighted')
f1 = f1_score(pred, y_test, average='weighted')

print('Recall : ', re)
```

```
print('Precision : ',pr)
print('F1 Score : ',f1)
```

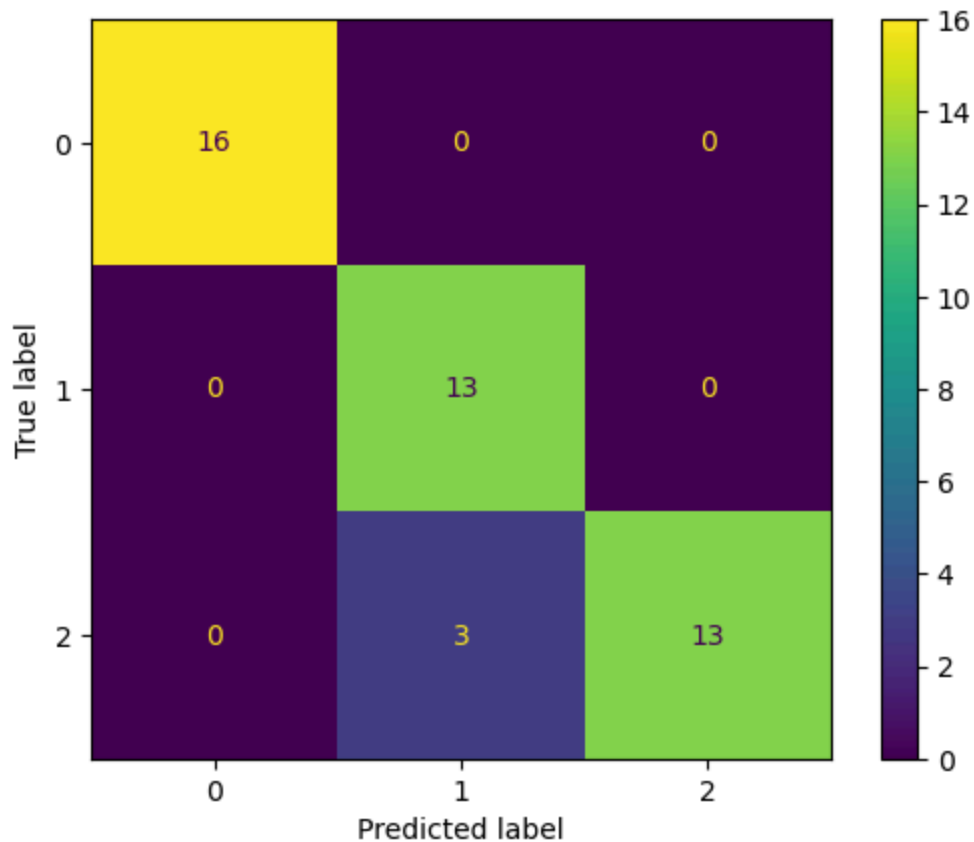
```
Accuracy : 0.9333333333333333
Recall : 0.9333333333333333
Precision : 0.9458333333333333
F1 Score : 0.9333333333333333
```

```
In [79]: from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
```

```
cm = confusion_matrix(pred,y_test)
cm
```

```
Out[79]: array([[16,  0,  0],
               [ 0, 13,  0],
               [ 0,  3, 13]], dtype=int64)
```

```
In [81]: cmd = ConfusionMatrixDisplay(cm)
cmd.plot()
plt.show()
```



```
In [87]: from sklearn.svm import SVC
svm = SVC(kernel='rbf', random_state=0, gamma=.10, C=1.0)
svm.fit(x_train, y_train)

svm.score(x_test, y_test)
```

```
Out[87]: 0.9333333333333333
```



```
In [89]: from sklearn.tree import DecisionTreeClassifier
dtree = DecisionTreeClassifier()
dtree.fit(x_train, y_train)

dtree.score(x_test, y_test)
```

Out[89]: 0.9333333333333333

```
In [91]: df = pd.DataFrame({'Y_Test':list(y_test),'Prediction':pred})
df
```

Out[91]:

	Y_Test	Prediction
0	0	0
1	1	2
2	1	1
3	2	2
4	2	2
5	0	0
6	0	0
7	2	2
8	2	2
9	1	1
10	2	2
11	1	1
12	0	0
13	2	2
14	2	2
15	2	2
16	0	0
17	1	1
18	2	2
19	0	0
20	0	0
21	2	2
22	1	2
23	0	0
24	0	0
25	1	1
26	0	0
27	0	0
28	0	0
29	1	1

	Y_Test	Prediction
30	0	0
31	1	1
32	0	0
33	1	1
34	1	1
35	1	1
36	1	2
37	2	2
38	2	2
39	1	1
40	0	0
41	0	0
42	2	2
43	1	1
44	1	1