

```
In [1]: ▶ import pandas as pd
from sklearn.datasets import load_iris
iris = load_iris()
```

```
In [2]: ▶ iris.feature_names
```

```
Out[2]: ['sepal length (cm)',
'sepal width (cm)',
'petal length (cm)',
'petal width (cm)']
```

```
In [3]: ▶ iris.target_names
```

```
Out[3]: array(['setosa', 'versicolor', 'virginica'], dtype='<U10')
```

```
In [4]: ▶ df = pd.DataFrame(iris.data, columns=iris.feature_names)
df.head()
```

```
Out[4]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
0	5.1	3.5	1.4	0.2
1	4.9	3.0	1.4	0.2
2	4.7	3.2	1.3	0.2
3	4.6	3.1	1.5	0.2
4	5.0	3.6	1.4	0.2

```
In [5]: ▶ df['target'] = iris.target
df.head()
```

```
Out[5]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target
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 [6]: ▶ df[df.target==1].head()
```

```
Out[6]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target
50	7.0	3.2	4.7	1.4	1
51	6.4	3.2	4.5	1.5	1
52	6.9	3.1	4.9	1.5	1
53	5.5	2.3	4.0	1.3	1
54	6.5	2.8	4.6	1.5	1

```
In [7]: ▶ df[df.target==2].head()
```

```
Out[7]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target
100	6.3	3.3	6.0	2.5	2
101	5.8	2.7	5.1	1.9	2
102	7.1	3.0	5.9	2.1	2
103	6.3	2.9	5.6	1.8	2
104	6.5	3.0	5.8	2.2	2

```
In [8]: df['flower_name'] =df.target.apply(lambda x: iris.target_names[x])
df.head()
```

Out[8]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower_name
0	5.1	3.5	1.4	0.2	0	setosa
1	4.9	3.0	1.4	0.2	0	setosa
2	4.7	3.2	1.3	0.2	0	setosa
3	4.6	3.1	1.5	0.2	0	setosa
4	5.0	3.6	1.4	0.2	0	setosa

```
In [9]: df[45:55]
```

Out[9]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower_name
45	4.8	3.0	1.4	0.3	0	setosa
46	5.1	3.8	1.6	0.2	0	setosa
47	4.6	3.2	1.4	0.2	0	setosa
48	5.3	3.7	1.5	0.2	0	setosa
49	5.0	3.3	1.4	0.2	0	setosa
50	7.0	3.2	4.7	1.4	1	versicolor
51	6.4	3.2	4.5	1.5	1	versicolor
52	6.9	3.1	4.9	1.5	1	versicolor
53	5.5	2.3	4.0	1.3	1	versicolor
54	6.5	2.8	4.6	1.5	1	versicolor

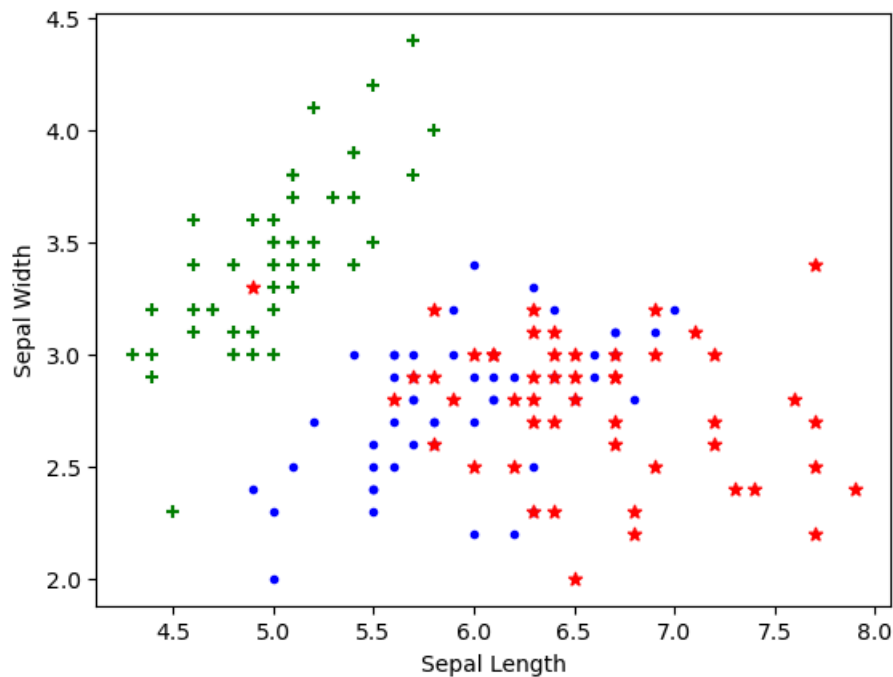
```
In [10]: df0 = df[:50]
df1 = df[50:100]
df2 = df[100:]
```

```
In [11]: import matplotlib.pyplot as plt
%matplotlib inline
```

## Sepal length vs Sepal Width (Setosa vs Versicolor vs virginica)

```
In [12]: ▶ plt.xlabel('Sepal Length')
plt.ylabel('Sepal Width')
plt.scatter(df0['sepal length (cm)'], df0['sepal width (cm)'],color="green",marker='+')
plt.scatter(df1['sepal length (cm)'], df1['sepal width (cm)'],color="blue",marker='.')
plt.scatter(df2['sepal length (cm)'], df1['sepal width (cm)'],color="red",marker='*')
```

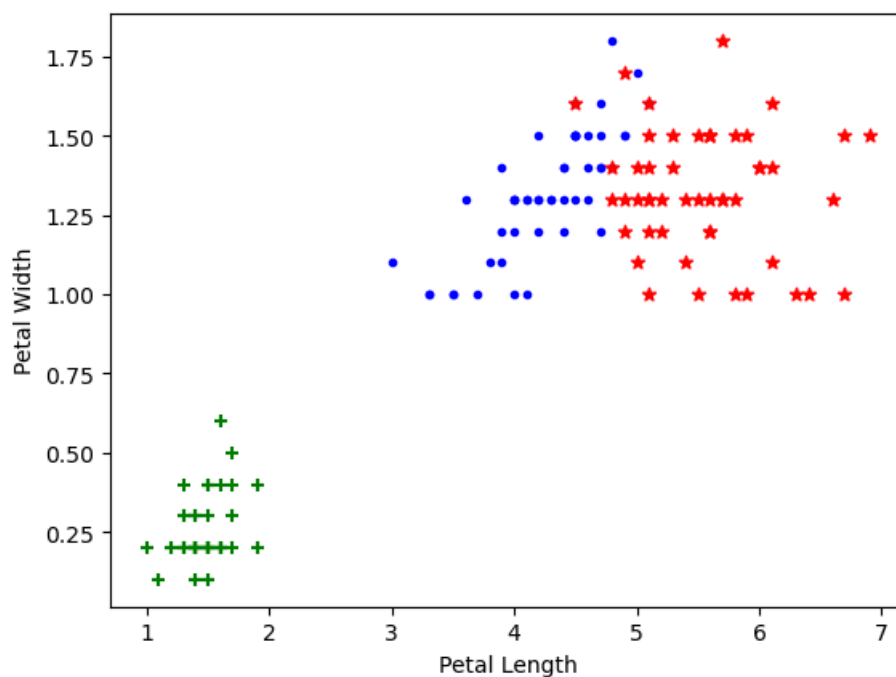
Out[12]: <matplotlib.collections.PathCollection at 0x221dd21bee0>



## Petal length vs Petal Width (Setosa vs Versicolor,virginica)

```
In [13]: ▶ plt.xlabel('Petal Length')
plt.ylabel('Petal Width')
plt.scatter(df0['petal length (cm)'], df0['petal width (cm)'],color="green",marker='+')
plt.scatter(df1['petal length (cm)'], df1['petal width (cm)'],color="blue",marker='.')
plt.scatter(df2['petal length (cm)'], df1['petal width (cm)'],color="red",marker='*')
```

Out[13]: <matplotlib.collections.PathCollection at 0x221ddb1e950>



## Train test split

```
In [14]: from sklearn.model_selection import train_test_split
```

```
In [15]: X = df.drop(['target', 'flower_name'], axis='columns')
y = df.target
```

```
In [16]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=1)
len(X_train)
```

```
Out[16]: 120
```

```
In [17]: len(X_test)
```

```
Out[17]: 30
```

## Create KNN (K Neihrest Neighbour Classifier)

```
In [18]: from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=10)
```

```
In [19]: knn.fit(X_train, y_train)
```

```
Out[19]: KNeighborsClassifier
KNeighborsClassifier(n_neighbors=10)
```

```
In [20]: knn.score(X_test, y_test)
```

```
Out[20]: 0.9666666666666667
```

```
In [21]: knn.predict([[4.8, 3.0, 1.5, 0.3]])
```

```
C:\Users\Asus\anaconda3\lib\site-packages\sklearn\base.py:420: UserWarning: X does not have valid feature names, but KNeighborsClassifier was fitted with feature names
warnings.warn(
```

```
Out[21]: array([0])
```

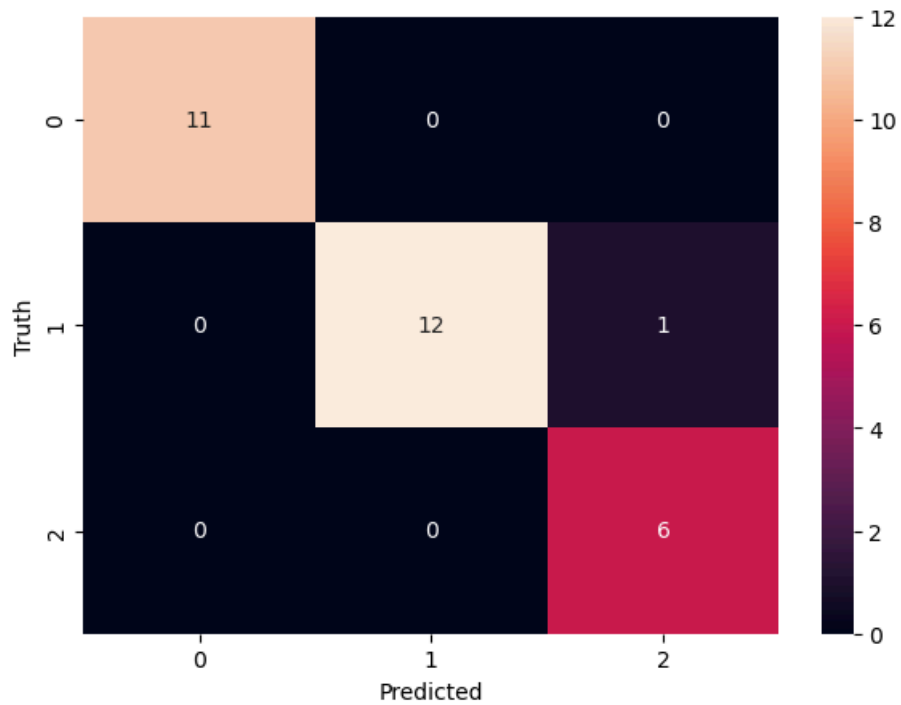
## Plot Confusion Matrix

```
In [22]: from sklearn.metrics import confusion_matrix
y_pred = knn.predict(X_test)
cm = confusion_matrix(y_test, y_pred)
cm
```

```
Out[22]: array([[11,  0,  0],
               [ 0, 12,  1],
               [ 0,  0,  6]], dtype=int64)
```

```
In [23]: ▶ %matplotlib inline
import matplotlib.pyplot as plt
import seaborn as sn
plt.figure(figsize=(7,5))
sn.heatmap(cm, annot=True)
plt.xlabel('Predicted')
plt.ylabel('Truth')
cm = confusion_matrix(y_test, y_pred)
cm
```

```
Out[23]: array([[11,  0,  0],
                [ 0, 12,  1],
                [ 0,  0,  6]], dtype=int64)
```



```
In [ ]: ▶
```