

KNN (K Nearest Neighbors) Classification

```
1 import pandas as pd
2 from sklearn.datasets import load_iris
3 iris = load_iris()
```

```
1 iris.feature_names
```

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

```
1 iris.target_names
```

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

```
1 df = pd.DataFrame(iris.data, columns=iris.feature_names)
2 df.head()
```

```
↗
```

	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	

```
1 df['target'] = iris.target
2 df.head()
```

	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

```
1 df[df.target==1].head()
```

	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

```
1 df[df.target==2].head()
```

	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

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

	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

```
1 df[45:55]
```

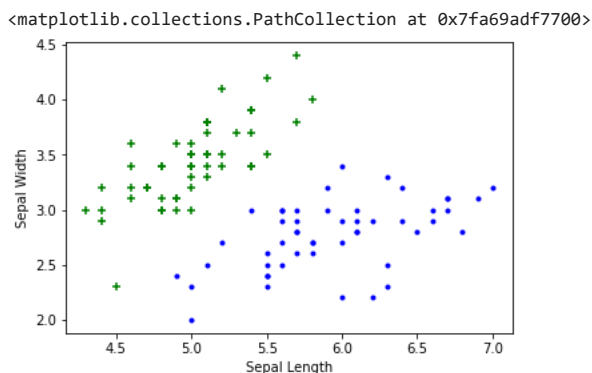
	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower_name
45	4.8	3.0	1.4	0.3	0	setosa
46	5.1	3.8	1.6	0.2	0	setosa
47	4.6	3.2	1.4	0.2	0	setosa
48	5.3	3.7	1.5	0.2	0	setosa
49	5.0	3.3	1.4	0.2	0	setosa
50	7.0	3.2	4.7	1.4	1	versicolor
51	6.4	3.2	4.5	1.5	1	versicolor
52	6.9	3.1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

```
1 df0 = df[:50]
2 df1 = df[50:100]
3 df2 = df[100:]
```

```
1 import matplotlib.pyplot as plt
2 %matplotlib inline
```

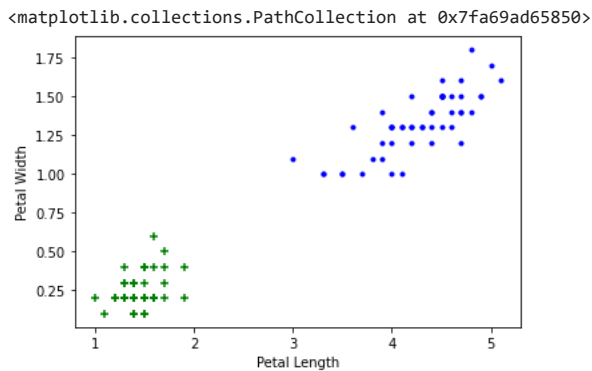
Sepal length vs Sepal Width (Setosa vs Versicolor)

```
1 plt.xlabel('Sepal Length')
2 plt.ylabel('Sepal Width')
3 plt.scatter(df0['sepal length (cm)'], df0['sepal width (cm)'],color="green",marker='+')
4 plt.scatter(df1['sepal length (cm)'], df1['sepal width (cm)'],color="blue",marker='.')
```



Petal length vs Petal Width (Setosa vs Versicolor)

```
1 plt.xlabel('Petal Length')
2 plt.ylabel('Petal Width')
3 plt.scatter(df0['petal length (cm)'], df0['petal width (cm)'],color="green",marker='+')
4 plt.scatter(df1['petal length (cm)'], df1['petal width (cm)'],color="blue",marker='.')
```



Train test split

```
1 from sklearn.model_selection import train_test_split

1 X = df.drop(['target', 'flower_name'], axis='columns')
2 y = df.target
3 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=1)

1 len(X_train)

120

1 len(X_test)

30
```

Create KNN (K Neighrest Neighbour Classifier)

```
1 from sklearn.neighbors import KNeighborsClassifier
2 knn = KNeighborsClassifier(n_neighbors=10)

1 knn.fit(X_train, y_train)

KNeighborsClassifier(n_neighbors=10)

1 knn.score(X_test, y_test)

0.9666666666666667

1 knn.predict([[4.8, 3.0, 1.5, 0.3]])

/usr/local/lib/python3.8/dist-packages/sklearn/base.py:450: UserWarning: X does not have valid feature names, but KNeighborsClassifier w
  warnings.warn(
array([0])
```

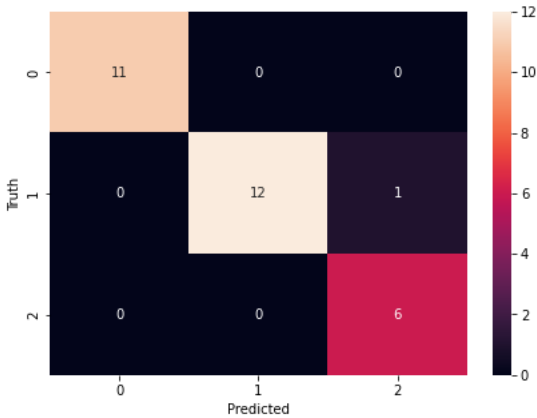
Plot Confusion Matrix

```
1 from sklearn.metrics import confusion_matrix
2 y_pred = knn.predict(X_test)
3 cm = confusion_matrix(y_test, y_pred)
4 cm

array([[11,  0,  0],
       [ 0, 12,  1],
       [ 0,  0,  6]])

1 %matplotlib inline
2 import matplotlib.pyplot as plt
3 import seaborn as sn
4 plt.figure(figsize=(7,5))
5 sn.heatmap(cm, annot=True)
```

```
6 plt.xlabel('Predicted')
7 plt.ylabel('Truth')
8 plt.title('Confusion Matrix')
```



```
1 from sklearn.metrics import classification_report
2
3 print(classification_report(y_test, y_pred))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	11
1	1.00	0.92	0.96	13
2	0.86	1.00	0.92	6
accuracy			0.97	30
macro avg	0.95	0.97	0.96	30
weighted avg	0.97	0.97	0.97	30

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