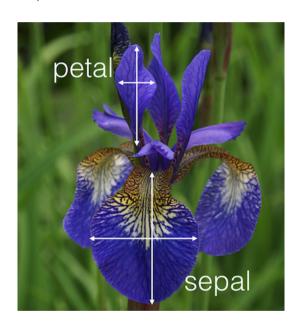
## KNN (K Nearest Neighbors) Classification

```
x
    array([0])

## Loading the necessary packages
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
from sklearn.datasets import load_iris
iris = load_iris()
%matplotlib inline
```



```
## Displaying the feature names
iris.feature_names

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

## Displaying the class labels
iris.target_names
```

array(['setosa', 'versicolor', 'virginica'], dtype='<U10')</pre>

## Creating a dataframe out of the dataset loaded

df = pd.DataFrame(iris.data,columns=iris.feature\_names)
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

## Adding class labels to the dataframe

df['target'] = iris.target
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	ılı
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	

df['flower\_name'] = df.target.apply(lambda x: iris.target\_names[x])
df.head()

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower_name	<b>=</b>
0	5.1	3.5	1.4	0.2	0	setosa	ıl.
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	

df.tail()

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower_name	
145	6.7	3.0	5.2	2.3	2	virginica	ılı
146	6.3	2.5	5.0	1.9	2	virginica	
147	6.5	3.0	5.2	2.0	2	virginica	

## Creating three dataframes corresponding to three class labels from the main dataframe df

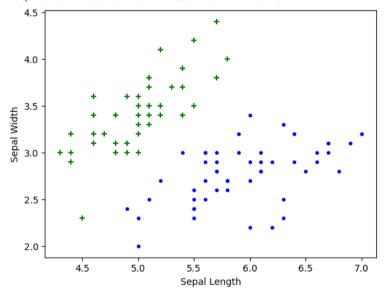
```
df0 = df[:50]
df1 = df[50:100]
df2 = df[100:]
```

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

```
## Visulaizing the instances of Setosa and Versicolor wrt. Sepal Length & Width
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='.')

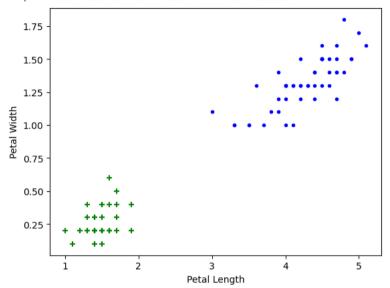
<matplotlib.collections.PathCollection at 0x7e1628b84e20>



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

## Visulaizing the instances of Setosa and Versicolor wrt. Sepal Length & Width
plt.xlabel('Petal Length')
plt.ylabel('Petal Width')

<matplotlib.collections.PathCollection at 0x7e1628c0ab60>



#### Train test split

```
from sklearn.model_selection import train_test_split

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

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=1)

print(len(X_train), len(X_test))

120 30
```

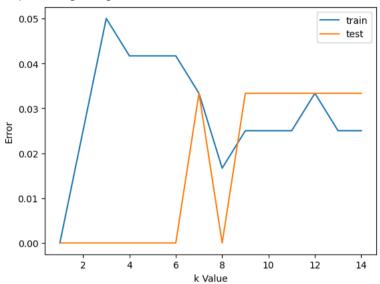
#### **Create KNN (K Neighrest Neighbour Classifier)**

```
## Finetuning the hyperparameter K
from sklearn.neighbors import KNeighborsClassifier
error1= []
error2= []
for k in range(1,15):
    knn= KNeighborsClassifier(n_neighbors=k)
    knn.fit(X_train,y_train)
```

```
y_pred1= knn.predict(X_train)
error1.append(np.mean(y_train!= y_pred1))
y_pred2= knn.predict(X_test)
error2.append(np.mean(y_test!= y_pred2))

# plt.figure(figsize(20,10))
plt.plot(range(1,15),error1,label="train")
plt.plot(range(1,15),error2,label="test")
plt.xlabel('k Value')
plt.ylabel('Error')
plt.legend()
```

<matplotlib.legend.Legend at 0x7e1628e06560>



```
## Model Creation
```

knn = KNeighborsClassifier(n\_neighbors=8)

## Model Training

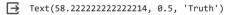
knn.fit(X\_train, y\_train)

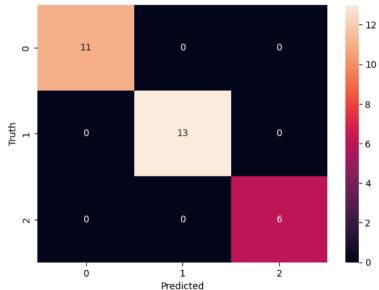
```
KNeighborsClassifier
KNeighborsClassifier(n_neighbors=8)
```

knn.score(X\_test, y\_test)

#### **Plot Confusion Matrix**

import 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')





Print classification report for precesion, recall and f1-score for each classes

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

	precision	recall	f1-score	support
0	1.00	1.00	1.00	11
1	1.00	1.00	1.00	13
2	1.00	1.00	1.00	6
accuracy			1.00	30
macro avg	1.00	1.00	1.00	30
weighted avg	1.00	1.00	1.00	30

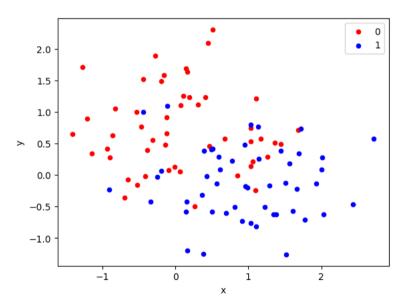
## **Generating Datasets**

#### **Blobs Classification Problem**

```
from sklearn.datasets import make_blobs
from matplotlib import pyplot
from pandas import DataFrame

# generate 2d classification dataset
X, y = make_blobs(n_samples=100, centers=2, n_features=2)

# scatter plot, dots colored by class value
df = DataFrame(dict(x1=X[:,0], x2=X[:,1], label=y))
colors = {0:'red', 1:'blue'}
fig, ax = pyplot.subplots()
grouped = df.groupby('label')
for key, group in grouped:
    group.plot(ax=ax, kind='scatter', x='x1', y='x2', label=key, color=colors[key])
pyplot.show()
```

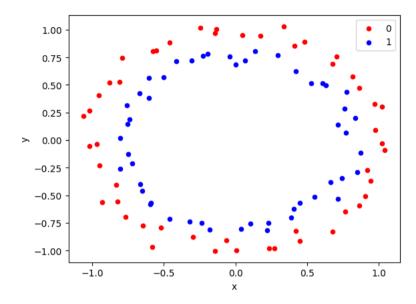


### **Circles Classification Problem**

```
from sklearn.datasets import make_circles

# generate 2d classification dataset
X, y = make_circles(n_samples=100, noise=0.05)
```

```
# scatter plot, dots colored by class value
df = DataFrame(dict(x=X[:,0], y=X[:,1], label=y))
colors = {0:'red', 1:'blue'}
fig, ax = pyplot.subplots()
grouped = df.groupby('label')
for key, group in grouped:
    group.plot(ax=ax, kind='scatter', x='x', y='y', label=key, color=colors[key])
```



# **Regression Test Problems**

```
from sklearn.datasets import make_regression

# generate regression dataset
X, y = make_regression(n_samples=100, n_features=1, noise=20)

# plot regression dataset
pyplot.scatter(X,y)
pyplot.show()
```

