AIM:

To write a Python program to calculate the KNN Algorithm by using the given series of files.

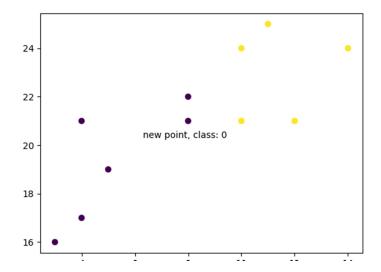
K-nearest neighbors (KNN) is a simple, supervised machine learning (ML) algorithm that can be used for classification or regression tasks - and is also frequently used in missing value imputation. It is based on the idea that the observations closest to a given data point are the most "similar" observations in a data set, and we can therefore classify unforeseen points based on the values of the closest existing points. By choosing *K*, the user can select the number of nearby observations to use in the algorithm.

Here, we will show you how to implement the KNN algorithm for classification, and show how different values of *K* affect the results.

K is the number of nearest neighbors to use. For classification, a majority vote is used to determined which class a new observation should fall into. Larger values of K are often more robust to outliers and produce more stable decision boundaries than very small values (K=3 would be better than K=1, which might produce undesirable results.

Program:

```
import matplotlib.pyplot as plt
from sklearn.neighbors import KNeighborsClassifier
x = [4, 5, 10, 4, 3, 11, 14, 8, 10, 12]
y = [21, 19, 24, 17, 16, 25, 24, 22, 21, 21]
classes = [0, 0, 1, 0, 0, 1, 1, 0, 1, 1]
data = list(zip(x, y))
knn = KNeighborsClassifier(n_neighbors=1)
knn.fit(data, classes)
new_x = 8
new y = 21
new_point = [(new_x, new_y)]
prediction = knn.predict(new point)
plt.scatter(x + [new_x], y + [new_y], c=classes + [prediction[0]])
plt.text(x=new x-1.7, y=new y-0.7, s=f"new point, class: {prediction[0]}")
plt.show()
Output:
```



(ii) Program:

```
import matplotlib.pyplot as plt from sklearn.neighbors import KNeighborsClassifier x = [4, 5, 10, 4, 3, 11, 14, 8, 10, 12] y = [21, 19, 24, 17, 16, 25, 24, 22, 21, 21] classes = [0, 0, 1, 0, 0, 1, 1, 0, 1, 1] data = list(zip(x, y)) knn = KNeighborsClassifier(n_neighbors=5) knn.fit(data, classes) new_x = 8 new_y = 21 new_point = [(\text{new}_x, \text{new}_y)] prediction = knn.predict(new_point) plt.scatter(x + [\text{new}_x], y + [\text{new}_y], c=classes + [\text{prediction}[0]]) plt.text(x=new_x-1.7, y=new_y-0.7, s=f"new point, class: {prediction}[0]}") plt.show()
```

Output:

