

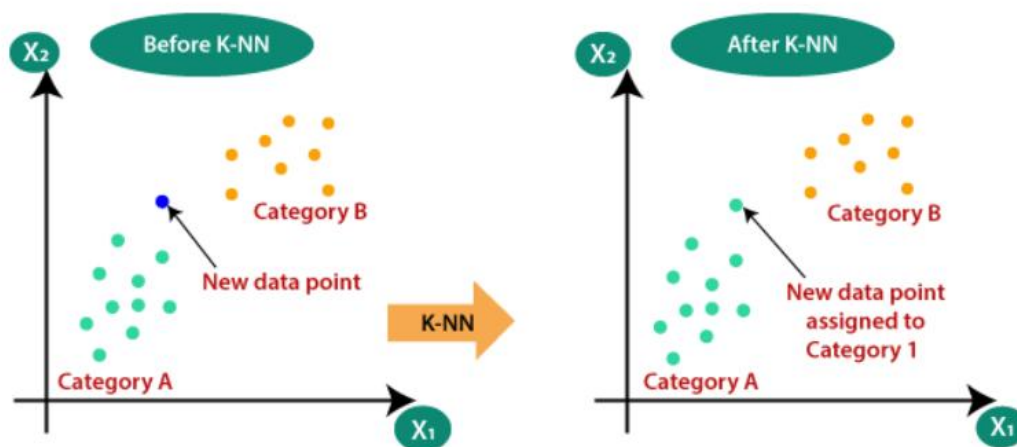
# K-Nearest Neighbors (KNN) Classifier

K-Nearest Neighbors (KNN) is a **simple and powerful** machine learning algorithm used for **classification and regression**.

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## How KNN Works?

1. **Choose a value of k** (number of neighbors).
  2. **Find the k nearest data points** based on distance (e.g., Euclidean distance).
  3. **Majority voting** (for classification) → The most common class among the k neighbors is assigned to the new data point.
  4. **Averaging** (for regression) → The output is the average of k nearest values.
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The K-NN working can be explained on the basis of the below algorithm:

- **Step-1:** Select the number K of the neighbors
- **Step-2:** Calculate the Euclidean distance of **K number of neighbors**
- **Step-3:** Take the K nearest neighbors as per the calculated Euclidean distance.
- **Step-4:** Among these k neighbors, count the number of the data points in each category.
- **Step-5:** Assign the new data points to that category for which the number of the neighbor is maximum.

- **Step-6:** Our model is ready.

### Implementing KNN in Python

```
from sklearn.neighbors import KNeighborsClassifier

# Create KNN model with k=5
knn = KNeighborsClassifier(n_neighbors=5)

# Train the model
knn.fit(trainData, trainLabel)

# Make predictions
predictions = knn.predict(testData)

# Evaluate the model
from sklearn.metrics import classification_report
print(classification_report(testLabel, predictions))
```

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### Choosing the Best k Value

- **Small k (e.g., 1-3)** → Can lead to **overfitting** (too sensitive to noise).
  - **Large k (e.g., 10-20)** → Leads to **smoother decision boundaries**, reducing variance.
  - **Common choice:**  $k = \sqrt{n}$ , where n is the number of training samples.
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### Advantages of KNN

- ✓ **Simple & Easy to implement**
- ✓ **Non-parametric** (doesn't assume a specific data distribution)
- ✓ **Works well for small datasets**

### Disadvantages of KNN

- ✗ **Computationally expensive** (slow on large datasets)
  - ✗ **Sensitive to irrelevant features** (feature scaling is necessary)
  - ✗ **Doesn't work well with high-dimensional data**
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Ref: <https://www.tpointtech.com/k-nearest-neighbor-algorithm-for-machine-learning>