

## ★ KNN Alg<sup>o</sup> :-

K-Nearest Neighbors (KNN) is a simple, non-parametric, and lazy machine learning algorithm used for classification and regression tasks.

It majorly works on finding the 'K' closest data points to a given point and making predictions based on these neighbours.

⇒ steps :-

01 Choose the number 'K' : determining the neighbors to consider when to classify a new data point.

02 Calculate the distance b/w the new data point

03 Sort the distance and select the 'K' nearest neighbors.

04 At last, the output.

05 Return the predicted label or value.



$$d = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2} \quad \{\text{Distance Metric}\}$$

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## # Code Using sklearn

```
iris = load_iris()
```

```
x = iris.data
```

```
y = iris.target
```

```
x_train, x_test, y_train, y_test = train_test_split(
    x, y, test_size=0.2, random_state=42)
```

```
scaler = StandardScaler()
```

```
x_train_scaled = scaler.fit_transform(x_train)
```

```
x_test_scaled = scaler.transform(x_test)
```

```
knn = KNeighborsClassifier(n_neighbors=3)
```

```
knn.fit(x_train_scaled, y_train)
```

```
y_pred = knn.predict(x_test_scaled)
```

```
accuracy = accuracy_score(y_test, y_pred)
```

## => Output:-

Accuracy of KNN classifier: 1.00

Predictions: [1 0 2 1 1 0 1 2 1 2 0 0 0 0  
1 2 1 2 0 2 0 2 2 2 2 2 0]

True labels: [1 0 2 1 1 0 1 2 1 2 0 0 0 0  
1 2 1 2 0 2 0 2 2 2 0 0]

## # Tuning 'K':

"If K is too small, the model may be noisy and overfit the data (high variance)"



If  $K$  is too large, the model may be too simple and underfit the data.

A common practice is to use cross-validation to find the optimal value for  $K$ .

## ★ SVM {Support Vector Machine} :-

a powerful, supervised machine learning algorithm commonly used for classification and regression tasks.

→ Algorithm :-

01 Input: A dataset with labeled examples.

02 Output: A hyperplane that best separates the classes in the feature space.

03 Training:

Find the optimal hyperplane by maximizing the margin b/w classes.

For non-linearly separable data, apply kernel functions to transform the data into higher dimensions, where a hyperplane can be found.

⇒ classification:



1 → Yes  
0 → No

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Input: A set of labeled data points

Output: A class label based on the optimal hyperplane.

Goal: Maximize the margin between classes while minimizing misclassification.

#Code:-

For Example:

• Age (in years)

• Income (in thousands of dollars)

• Product Usage Frequency (scale from 1 to 10)

```
np.random.seed(42)
```

```
n_Samples = 1000
```

```
age = np.random.randint(18, 70, n_Samples)
```

```
Income = np.random.randint(30, 150, n_Samples)
```

```
usage_freq = np.random.randint(1, 11, n_Samples)
```

Output:-

Accuracy of SVM Classification on Customer Purchase

Prediction = 0.99

Prediction: [0 1 1 1 0 0 1 0]

True labels = 521

/c