

# K-Nearest Neighbors (KNN)

## Introduction

K-Nearest Neighbors (KNN) is a simple, non-parametric, and supervised learning algorithm used

for both classification and regression tasks. It works by finding the 'K' closest data points (neighbors) to a query point and making predictions based on majority voting (classification) or averaging (regression).

## How KNN Works

1. Choose the number of neighbors (K).
2. Calculate the distance between the query point and all training points.
3. Select the K nearest neighbors based on distance.
4. For classification: assign the majority class among neighbors. For regression: take the average of neighbor values.

## Distance Metrics

Common distance metrics include: - Euclidean Distance - Manhattan Distance - Minkowski Distance - Hamming Distance (for categorical variables)

## Choosing K

The value of K greatly influences performance: - Small K → more flexible but sensitive to noise (high variance). - Large K → more stable but can oversmooth decision boundaries (high bias).

## Advantages

- Simple and intuitive. - No training phase (lazy learner). - Works well with smaller datasets.

## Disadvantages

- Computationally expensive for large datasets. - Sensitive to irrelevant or scaled features. - Performance drops in high-dimensional data (curse of dimensionality).

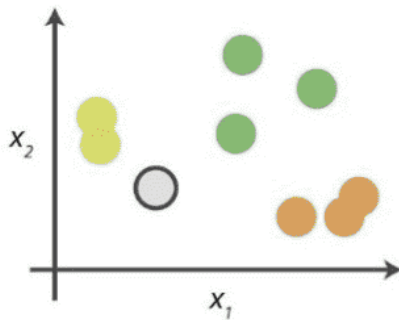
## Applications

- Recommender systems - Image recognition - Medical diagnosis - Anomaly detection

## Conclusion

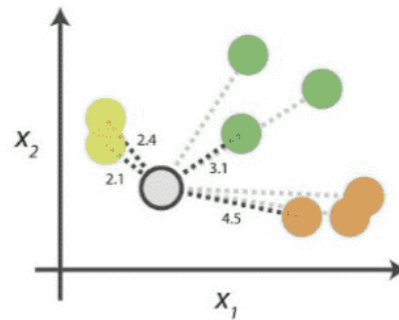
KNN is a powerful yet simple algorithm suitable for various applications. Its effectiveness depends on the choice of K, distance metric, and proper feature scaling.

## 0. Look at the data











Say you want to classify the grey point into a class. Here, there are three potential classes - lime green, green and orange.

## 1. Calculate distances









Start by calculating the distances between the grey point and all other points.

## 2. Find neighbours

Point Distance		
 ... 	2.1	→ 1st NN
 ... 	2.4	→ 2nd NN
 ... 	3.1	→ 3rd NN
 ... 	4.5	→ 4th NN

Next, find the nearest neighbours by ranking points by increasing distance. The nearest neighbours (NNs) of the grey point are the ones closest in dataspace.

## 3. Vote on labels

Class	# of votes	
	2	→ Class  wins the vote! Point  is therefore predicted to be of class  .
	1	
	1	

Vote on the predicted class labels based on the classes of the  $k$  nearest neighbours. Here, the labels were predicted based on the  $k=3$  nearest neighbours.