

Welcome to

# INTERNSHIP STUDIO

Module 04 | Lesson 04

K nearest neighbour





### **Objective:**

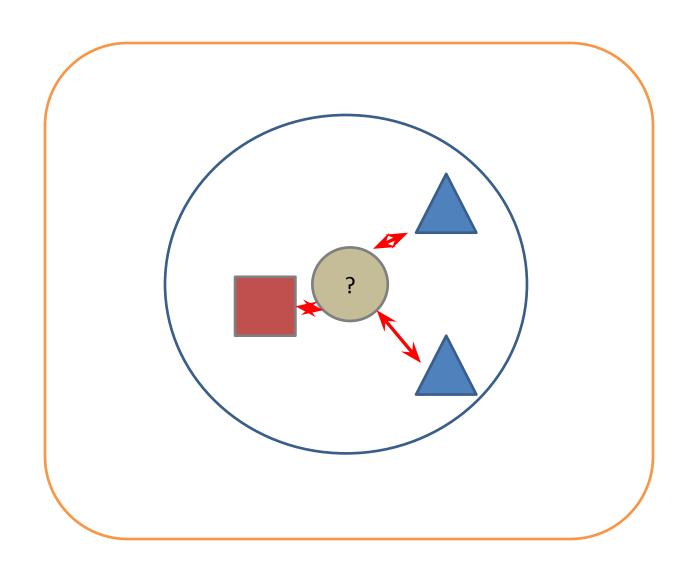
K Nearest Neighbor is an <u>instance based learning</u> <u>approach</u> that classifies <u>unlabeled data</u> to any of the classes based on their <u>similarity</u> and <u>majority voting</u>.



### Requirements:

- 1. The set of stored patterns.
- 2. Distance metric to compute distance between patterns.
- 3. The value of k (number of nearest neighbor)
  - Usually we take distance as Euclidian distance.
  - K should be any positive integer (odd numbers to avoid ties).
  - Attributes may have to be scaled to prevent distance measures from being dominated by one of the attributes

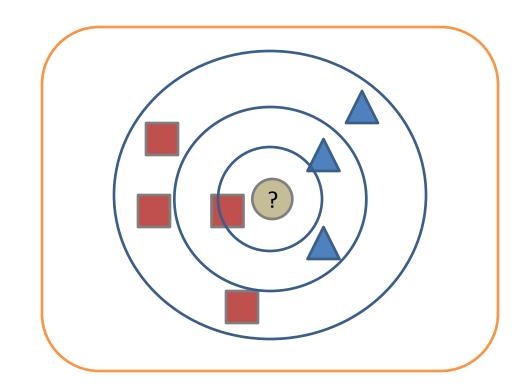




### **Classification Process:**

- 1. Take unlabeled data point.
- 2. Measure its distance from each of the class.
- 3. Use class labels of nearest neighbors to determine the class label of unknown pattern (e.g., by taking majority vote)





$$\Box k = 1$$
:

Belongs to square class

$$\Box k = 3$$
:

Belongs to triangle class

$$\Box k = 7$$
:

Belongs to square class

#### Choosing the value of *k* :

- If *k* is too small, sensitive to noise points
- lacktriangle If k is too large, neighborhood may include points from other classes



### Advantages

- ✓ Doesn't require apriori probability and distribution function.
- Easy to implement.

### Disadvantages

- ✓ If there are n elements complexity will be O(n) (computationally extensive)
- ✓ Sometime majority can dominate the classification.
- ✓ Accuracy can be severely degraded by the presence of noisy or irrelevant features