# K-Nearest Neighbor (KNN) Algorithm

BY:

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#### Outline...

Defining The Instance Based Learning

- Types Of Instance Based Learning
- K-nearest Neighbors (KNN) Algorithm
- Example Of KNN In Two Dimensional Space
- Example Of Data Set To Apply The Knn Algorithm

#### **Instance Based Learning:**

Its simply store the training example.

Relationship of new query instance to the previously stored examples is aimed to assign a target function value for new instance.

It includes two methods:

Nearest Neighbor

Weighted Regression

It is assumed that instances can be represented as points in Euclidean space.

Its some times referred as "Lazy" learning method.

## K-Nearest Neighbor Learner (KNN)

- It is most basic instance-base method.
- It assumes that all instance correspond to the points in the n-dimensional Euclidean space.
- Let a instance space  $\chi$  can be described as feature vector

$$x = \{a_1(x), a_2(x), a_3(x), \dots, a_n(x)\}$$

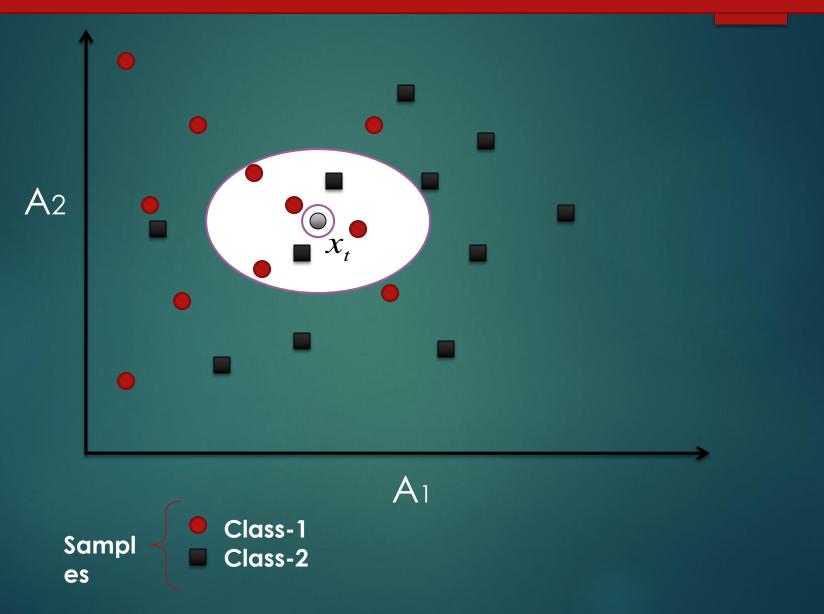
## K-Nearest Neighbor Learner (KNN)

• where  $a_i(x)$  denotes the value of the  $t^{th}$  attribute of instance x. Then the distance between instance  $X_i$  and  $X_j$ 

can be defined:

$$d(x_i, x_j) = \sqrt{\sum_{t=1}^{n} \{a_t(x_i) - a_t(x_j)\}^2 \dots (1)}$$

### **Two Dimensional Space**



### Distance-Weight Nearest Neighbor

The refinement of the k-Nearest Neighbor method is to assign weight corresponding their distance to the query point.

Then, above equation can be modified as:

$$\hat{f}(x_t) \leftarrow \arg\max_{y \in Y} \sum_{i=1}^k w_i \times \ell(y, f(x_i)).....(4)$$

#### Distance-Weight Nearest Neighbor

$$w_i = \frac{1}{d(x_t, x_i)^2}$$
 ..... (5)

The real valued target functions can also be modify in a similar manner, such as:

**Example:** Find out the labels of the given test sample using training set.

| S.N. | A1   | A2   | A3   | Labels |
|------|------|------|------|--------|
| D1   | 3.54 | 5.63 | 3.85 | 2.63   |
| D2   | 3.23 | 1.86 | 2.84 | 2.65   |
| D3   | 4.76 | 3.87 | 3.75 | 3.39   |
| D4   | 2.45 | 2.07 | 4.84 | 2.74   |
| D5   | 3.51 | 4.46 | 2.86 | 3.58   |
| Dt   | 3.98 | 8.97 | 3.75 | Ś      |

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Thank You