

# Machine Learning - Lecture k-Nearest-Neighbor classifier

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## k-Nearest-Neighbor classifier

## kNN classifier - An Illustrative Example



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# Euclidean Distance

The Euclidean distance between two points,  $x_1 = (x_{11}, x_{12}, \dots, x_{1n})$  and  $x_2 = (x_{21}, x_{22}, \dots, x_{2n})$ , is shown in Eq. 1

$$\text{dist}(x_1, x_2) = \sqrt{\sum_{i=1}^n (x_{1i} - x_{2i})^2} \quad (1)$$

The distance between the two points in the plane with coordinate  $(x,y)$  and  $(a,b)$  is given by:

$$ManhattanDistance, (x, y)(a, b) = |x - a| + |y - b| \quad (3)$$



Figure: Euclidean and Manhattan distance.

## kNN classifier (con.)

- ▶ For, kNN classifier, the unknown instance,  $x_{unknown}$  is assigned the most common class,  $c_l$  among its  $k$  nearest neighbours.
- ▶ The  $k$  is chosen to be odd for a two class classification and in general not to be a multiple of the number of classes  $M$ .
- ▶ Usually, kNN achieves good results when the data set is large.
- ▶ The value of  $k$  should be large for classifying the noisy data.
- ▶ Also we can consider the majority voting over the  $k$  nearest neighbours to deal with noisy instances.
- ▶ Algorithm 1 outlines the k-nearest-neighbor algorithm.

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**Algorithm 1** k-nearest-neighbor classifier

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**Input:**  $D = \{x_1, \dots, x_i, \dots, x_n\}$

**Output:** kNN classifier,  $kNN$ .

**Method:**

- 1: find  $X \in D$  that identify the  $k$  nearest neighbours, *regardless* of class label,  $c_l$ .
  - 2: out of these instances,  $X = \{x_1, x_2, \dots, x_k\}$ , identify the number of instances,  $k_i$ , that belong to class  $c_l$ ,  $l = 1, 2, \dots, M$ . Obviously,  $\sum_i k_i = k$ .
  - 3: assign  $x_{test}$  to the class  $c_l$  with the maximum number of  $k_i$  of instances.
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## Disadvantage of kNN classifier

- ▶ The main disadvantage of the kNN classifier is that it is a lazy learner, i.e. it does not learn anything from the training data and simply uses the training data itself for classification.
- ▶ A serious drawback associated with ( $k$ )NN technique is the complexity,  $(O(kN))^2$ , in search of the nearest neighbour(s) among the  $N$  available training samples. Although, due to its asymptotic error performance, the  $k$ NN rule achieves good results when the data set is large, the performance of the classifier may degrade dramatically when the value of  $N$  training instances is relatively small.

# An Illustrative Example

**Table:** Data for Height Classification.

Name	Gender	Height	Output
Kristina	F	1.6 m	Short
Jim	M	2 m	Tall
Maggie	F	1.9 m	Medium
Martha	F	1.88 m	Medium
Stephanie	F	1.7 m	Short
Bob	M	1.85 m	Medium
Kathy	F	1.6 m	Short
Dave	M	1.7 m	Short
Worth	M	2.2 m	Tall
Steven	M	2.1 m	Tall
Debbie	F	1.8 m	Medium
Todd	M	1.95 m	Medium
Kim	F	1.9 m	Medium
Amy	F	1.8 m	Medium
Wynette	F	1.75 m	Medium

## An Illustrative Example (con.)

- ▶ Using the sample data from Table 1 and the **Output** classification as the training set output value, we classify the instance (**Pat, F, 1.6**).
- ▶ Only the height is used for distance calculation so that both the Euclidean and Manhattan distance measures yield the same results; that is, the distance is simply the absolute value of the difference between the values.
- ▶ Suppose that  $K = 5$  is given. We then have that the  $K$  nearest neighbours to the input instance are (**Kristina, F, 1.6**), (**Kathy, F, 1.6**), (**Stephanie, F, 1.7**), (**Dave, M, 1.7**), and (**Wynette, F, 1.75**).
- ▶ Of these the five item, four are classified as short and one as medium. Thus, the kNN will classify **Pat** as **short**.

\*\*\* THANK YOU \*\*\*

