

# Data Mining

## Classification: Alternative Techniques

Lecture Notes for Chapter 4

Instance-Based Learning

Introduction to Data Mining , 2<sup>nd</sup> Edition

by

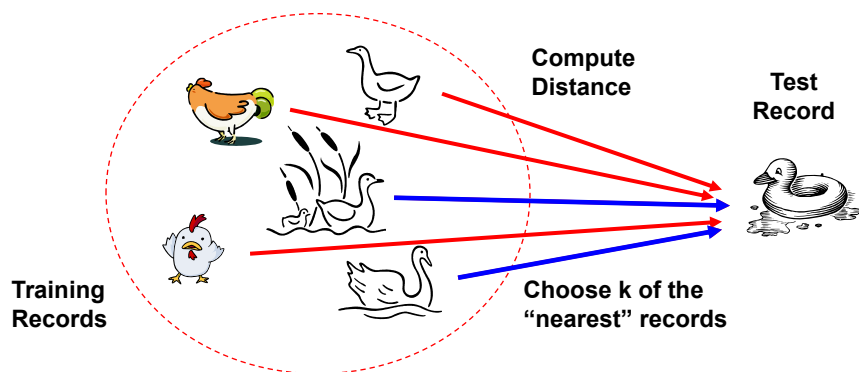
Tan, Steinbach, Karpatne, Kumar

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## Nearest Neighbor Classifiers

- Basic idea:

- If it walks like a duck, quacks like a duck, then it's probably a duck



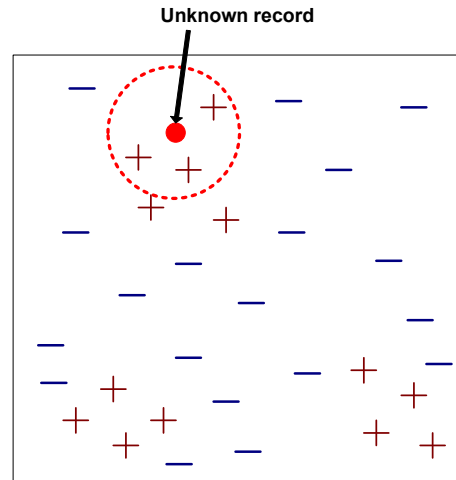
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## Nearest-Neighbor Classifiers



- Requires the following:
  - A set of labeled records
  - Proximity metric to compute distance/similarity between a pair of records
    - e.g., Euclidean distance
  - The value of  $k$ , the number of nearest neighbors to retrieve
  - A method for using class labels of  $K$  nearest neighbors to determine the class label of unknown record (e.g., by taking majority vote)

## How to Determine the class label of a Test Sample?

- Take the majority vote of class labels among the  $k$ -nearest neighbors
- Weight the vote according to distance
  - weight factor,  $w = 1/d^2$

## Choice of proximity measure matters

- For documents, cosine is better than correlation or Euclidean

1 1 1 1 1 1 1 1 1 1 0	VS	0 0 0 0 0 0 0 0 0 0 1
0 1 1 1 1 1 1 1 1 1 1		1 0 0 0 0 0 0 0 0 0 0

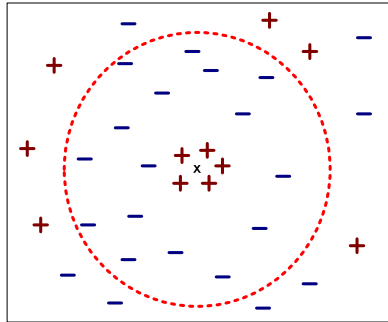
Euclidean distance = 1.4142 for both pairs, but the cosine similarity measure has different values for these pairs.

## Nearest Neighbor Classification...

- **Data preprocessing is often required**
  - Attributes may have to be scaled to prevent distance measures from being dominated by one of the attributes
    - ◆ Example:
      - height of a person may vary from 1.5m to 1.8m
      - weight of a person may vary from 90lb to 300lb
      - income of a person may vary from \$10K to \$1M
  - Time series are often standardized to have 0 means a standard deviation of 1

## Nearest Neighbor Classification...

- Choosing the value of k:
  - If k is too small, sensitive to noise points
  - If k is too large, neighborhood may include points from other classes



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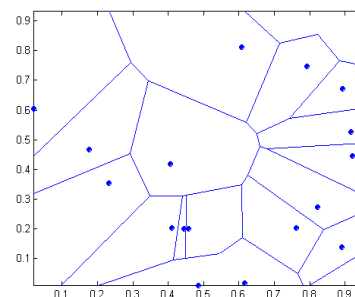
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## Nearest-neighbor classifiers

- Nearest neighbor classifiers are local classifiers
- They can produce decision boundaries of arbitrary shapes.

1-nn decision boundary is a Voronoi Diagram



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## Nearest Neighbor Classification...

### ● How to handle missing values in training and test sets?

- Proximity computations normally require the presence of all attributes
- Some approaches use the subset of attributes present in two instances
  - ◆ This may not produce good results since it effectively uses different proximity measures for each pair of instances
  - ◆ Thus, proximities are not comparable

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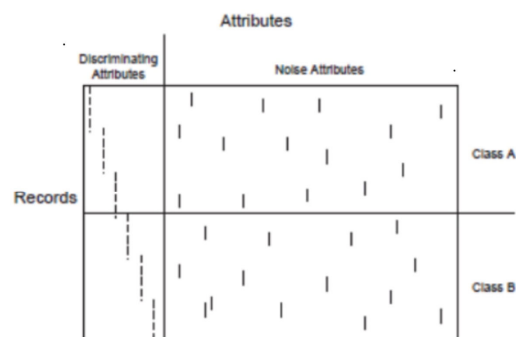
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## K-NN Classifiers...

### Handling Irrelevant and Redundant Attributes

- Irrelevant attributes add noise to the proximity measure
- Redundant attributes bias the proximity measure towards certain attributes



(a) Synthetic data set 1.

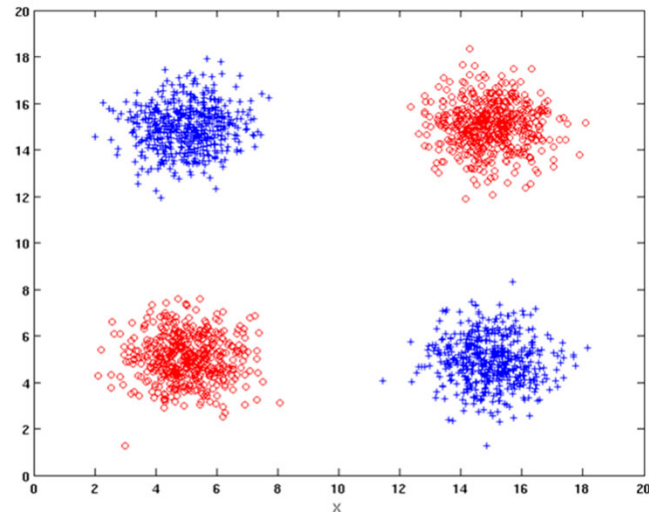
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## K-NN Classifiers: Handling attributes that are interacting



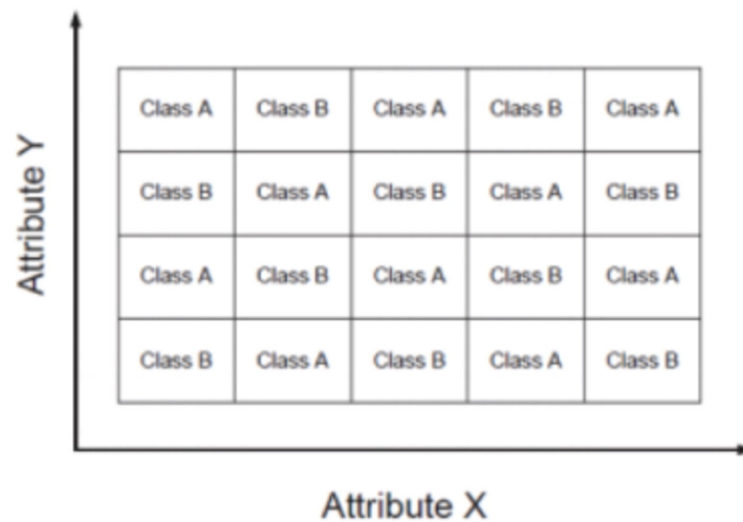
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## Handling attributes that are interacting



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## Improving KNN Efficiency

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- Avoid having to compute distance to all objects in the training set
  - Multi-dimensional access methods (k-d trees)
  - Fast approximate similarity search
  - Locality Sensitive Hashing (LSH)
- Condensing
  - Determine a smaller set of objects that give the same performance
- Editing
  - Remove objects to improve efficiency