



**Data Mining**

**Semi-supervised classification**

**Nearest Neighbors**

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**NOVA-IMS**

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**AGENDA**

- Cluster analysis
  - Semi-supervised classification
  - Nearest Neighbors

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## Semi-supervised classification

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UNIGIS  
AEGIS  
ISchools  
eduniversal

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### Going from clustering into classification

Unlabeled Data → Clustering - Labels → Classification – KNN and Trees

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## Going from clustering into classification



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## ***k*-nearest neighbors (*k*-NN)**

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## k-nearest neighbors

- Instance based classification:
  - Simplest form of learning;
  - Training instances are searched for instances that most closely resembles new instance;
  - The instances themselves represent the knowledge;
  - Also called instance-based learning
  - Similarity function defines what's "learned"

## k-nearest neighbors

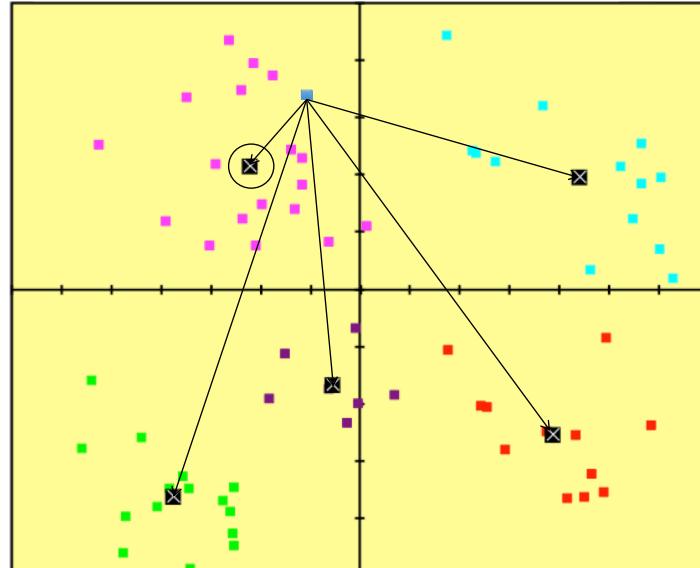
- Requires three things:
  1. The set of stored records (with labels)
  2. A distance metric to compute distance between records (can use Euclidean distance)
  3. The value of k, the number of nearest neighbors to retrieve

## *k*-nearest neighbors

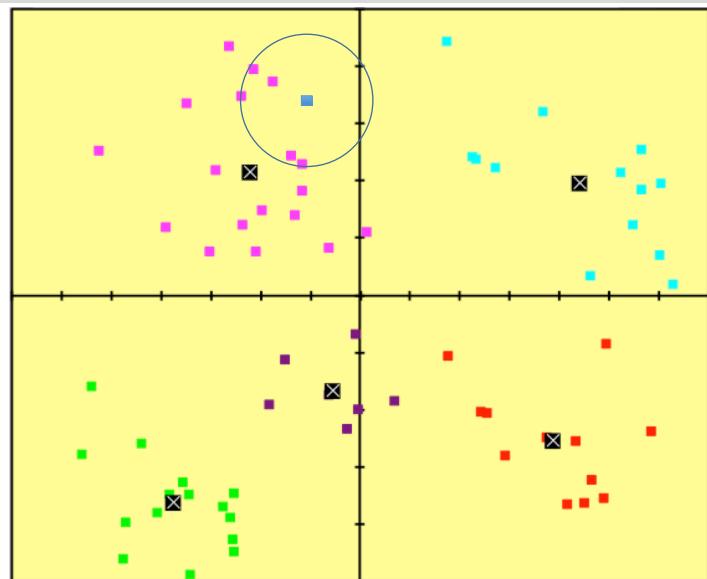
- To classify an unknown record:

1. Compute distance to other training records
2. Identify  $k$  nearest neighbors
3. Use class labels of nearest neighbors to determine the class label of unknown record (e.g., by taking majority vote)

## Going from clustering into classification

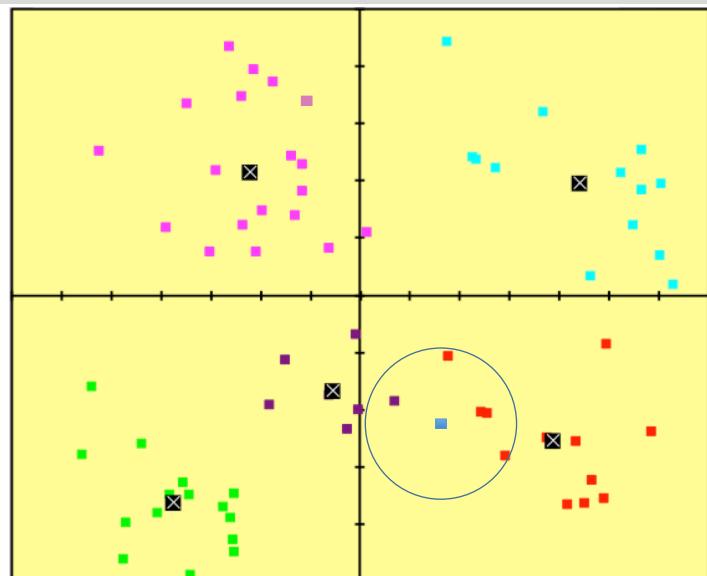


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## k-nearest neighbors

- Compute distance between two points:
    - Euclidean distance
- $$d(p, q) = \sqrt{\sum_i (p_i - q_i)^2}$$
- Determine the class from nearest neighbor list
    - Take the majority vote of class labels among the k-nearest neighbors
    - Weigh the vote according to distance
      - weight factor,  $w = 1/d^2$

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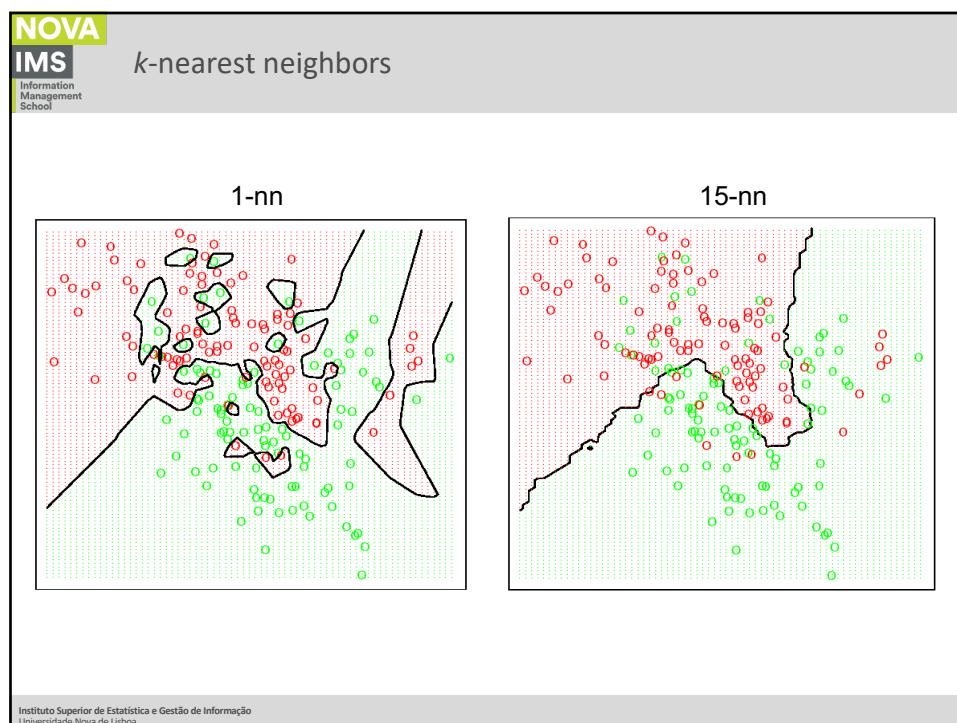
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## k-nearest neighbors

- k-nn frontiers (and the number  $k$ ):
  - Large  $k$ 
    - Smooth frontiers
    - Unable to detect small variations
  - Small  $k$ 
    - Very sensitive to outliers
    - Crisp frontiers

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