

Instance Based Classifiers

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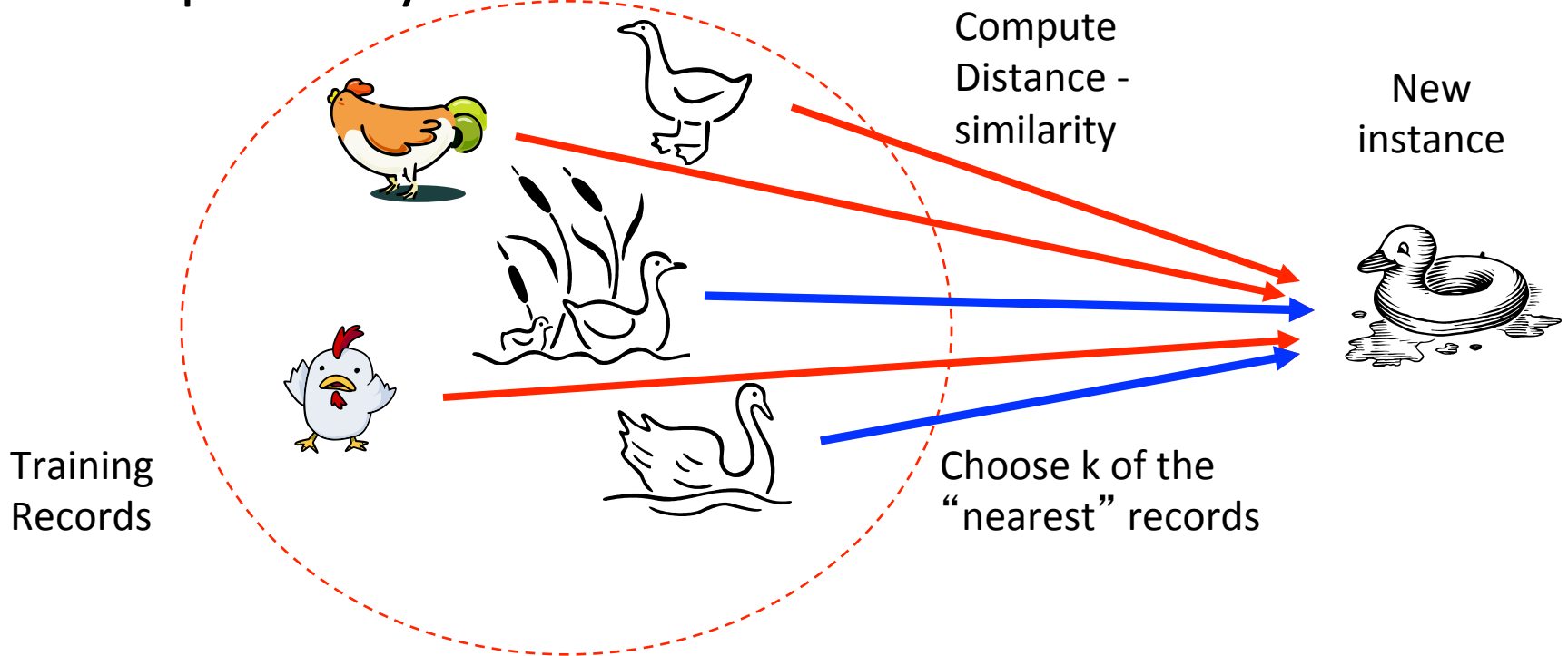
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Instance Based Classifiers

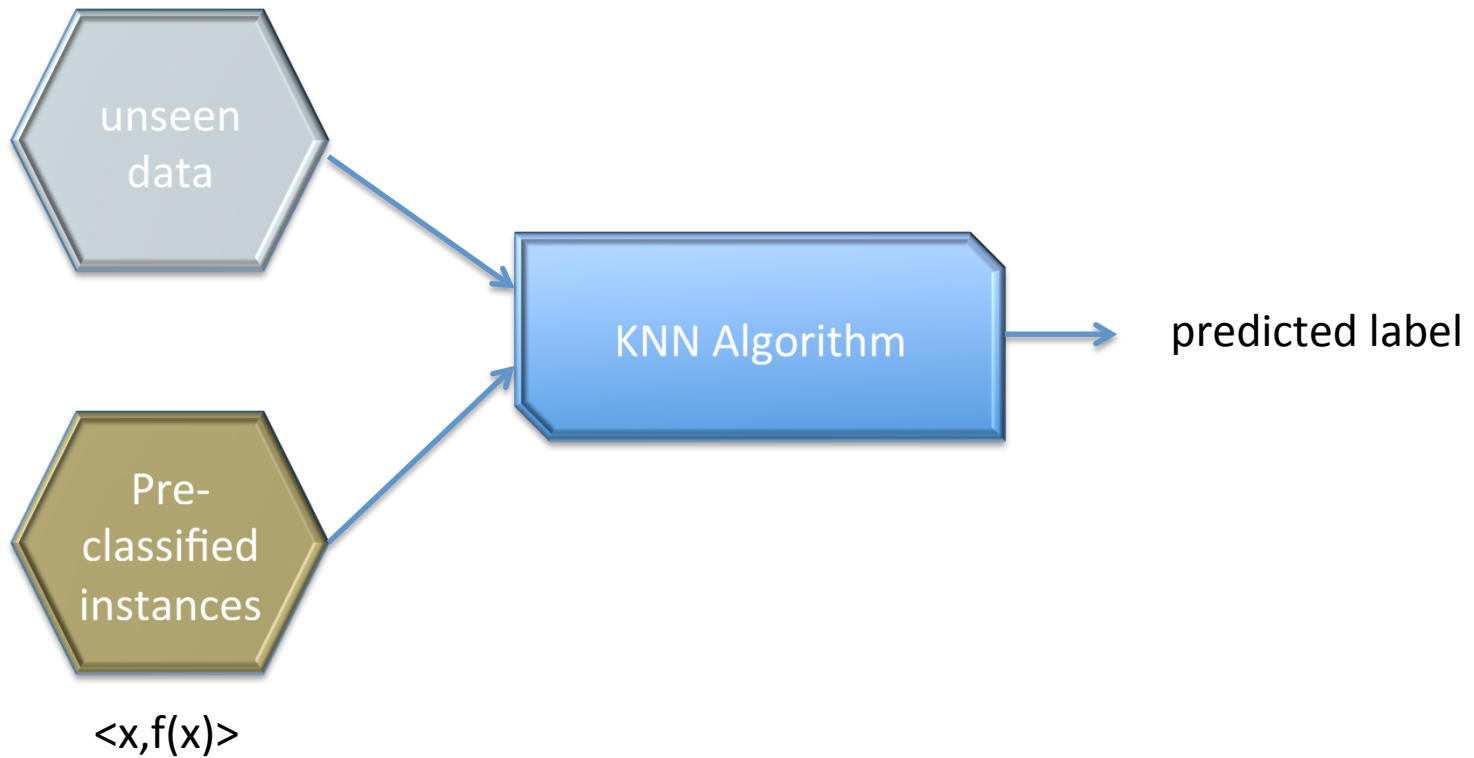
- Instance-based classifiers do not induce a model from training data
- On the contrary, they use a set of pre-classified instances to predict “on the fly” the class label of unseen cases
- K-Nearest Neighbors (KNN)

Nearest Neighbor Classifiers

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

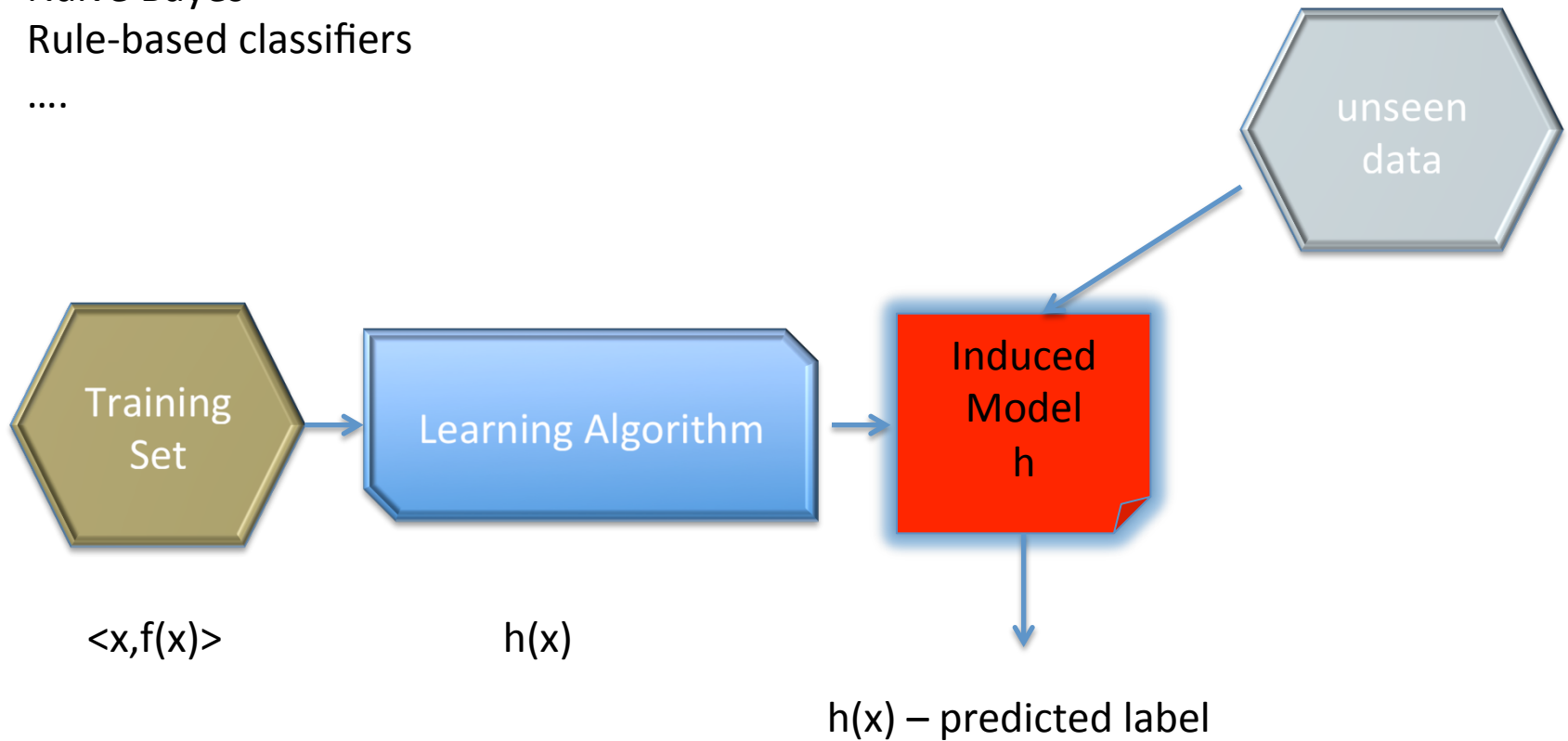


Lazy classifiers



Eager classifiers

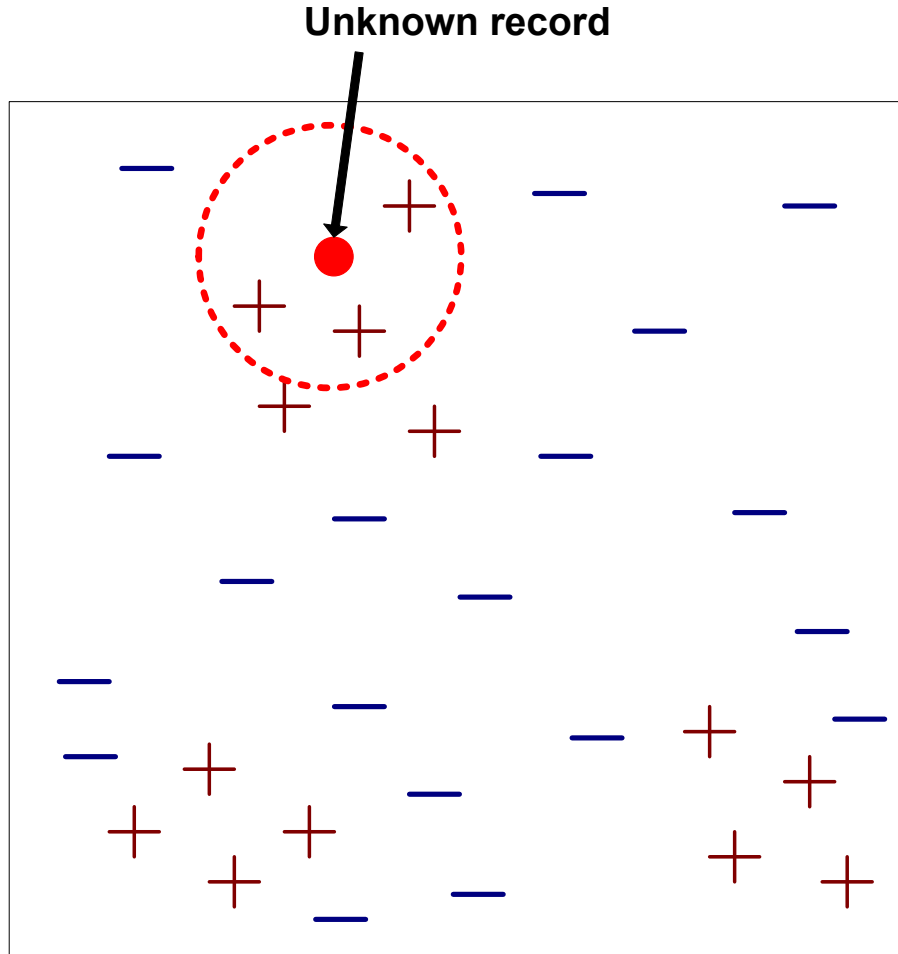
- Decision trees
- Classification rules
- Naïve Bayes
- Rule-based classifiers
-



K-Nearest Neighbors (kNN)

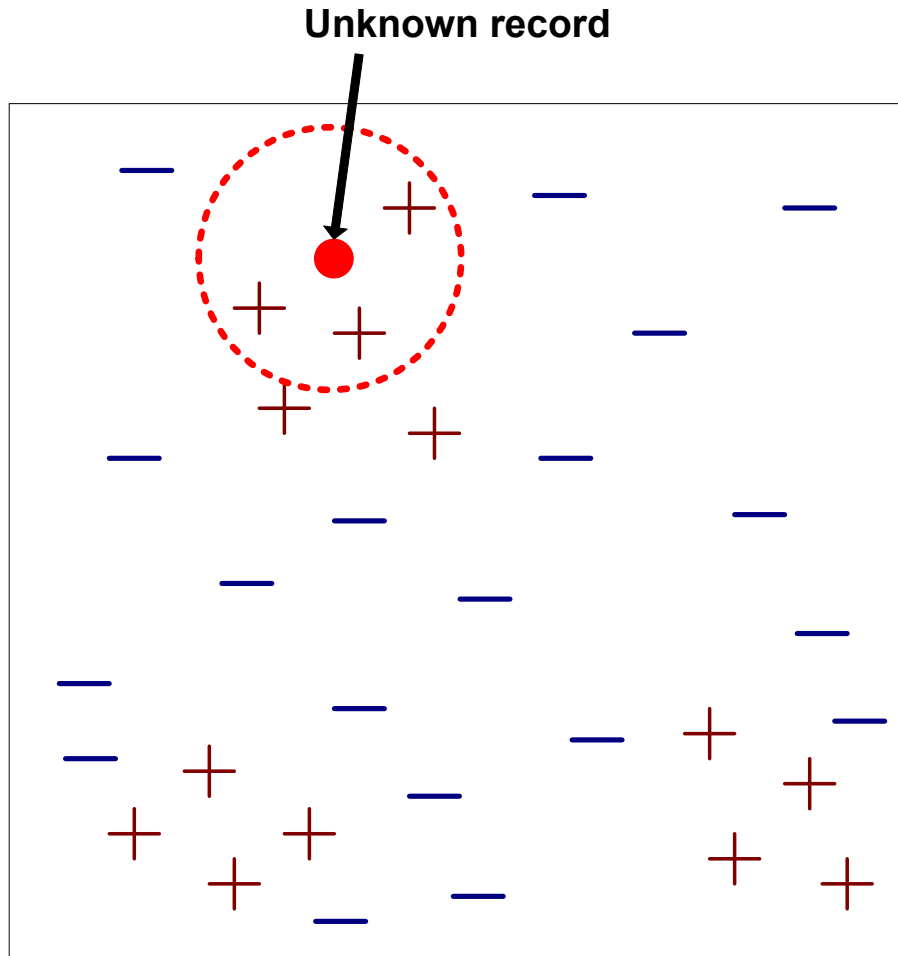
- K-Nearest Neighbors (kNN)
 - Uses k “closest” examples (nearest neighbors) in the training set for performing classification
 - “Closeness” is computed by using a Similarity function (or Distance function)

K-Nearest-Neighbor Classifiers



- Requires three things
 - The set of pre-classified instances
 - Distance Metric to compute distance between records
 - The value of k , the number of nearest neighbors to retrieve
- To classify an unseen instance X :
 - Compute distance of X to other instances
 - Identify k nearest neighbors (smallest distance, highest similarity)
 - Use class labels of k nearest neighbors to determine the class label of unseen instance (e.g., by taking majority vote)

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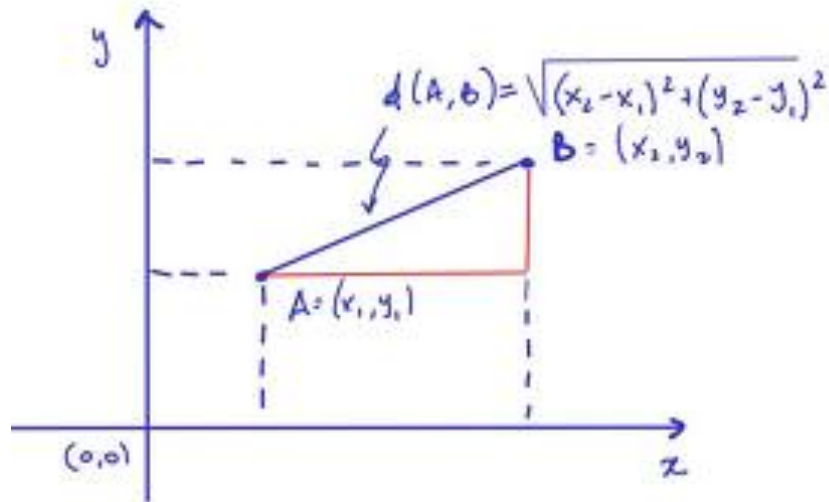
Compute distance of X to other instances - Euclidean distance

- Compute distance between two points:
 - Euclidean distance

$$d(p, q) = \sqrt{\sum_i^n (p_i - q_i)^2}$$

- where p and q are two instances, n is the number of their attributes, and p_i and q_i the values of the i-th attributes of p and q. Euclidean distances apply only to numerical attributes

Compute distance of X to other instances - Euclidean distance



$$\text{2-distanza} = \sqrt{\sum_{i=1}^n |x_i - y_i|^2}$$

Compute distance of X to other instances - SMC

- Simple Matching Coefficient:

$$- \text{SMC} = \frac{\text{number of matching attribute values}}{\text{Number of attributes}}$$

- Given

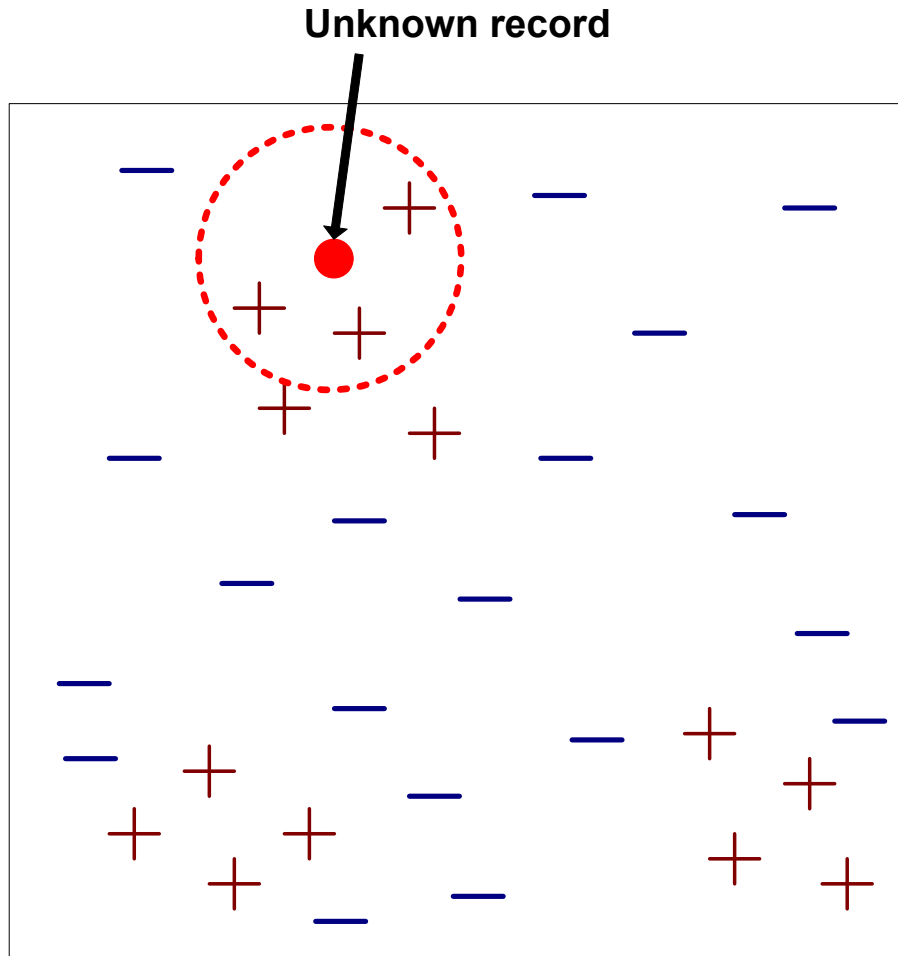
$X_1 = \langle 15, \text{rome}, \text{yellow} \rangle$

$X_2 = \langle 20, \text{paris}, \text{yellow} \rangle$

$$\text{SMC}(X_1, X_2) = 1/3 = 0.33$$

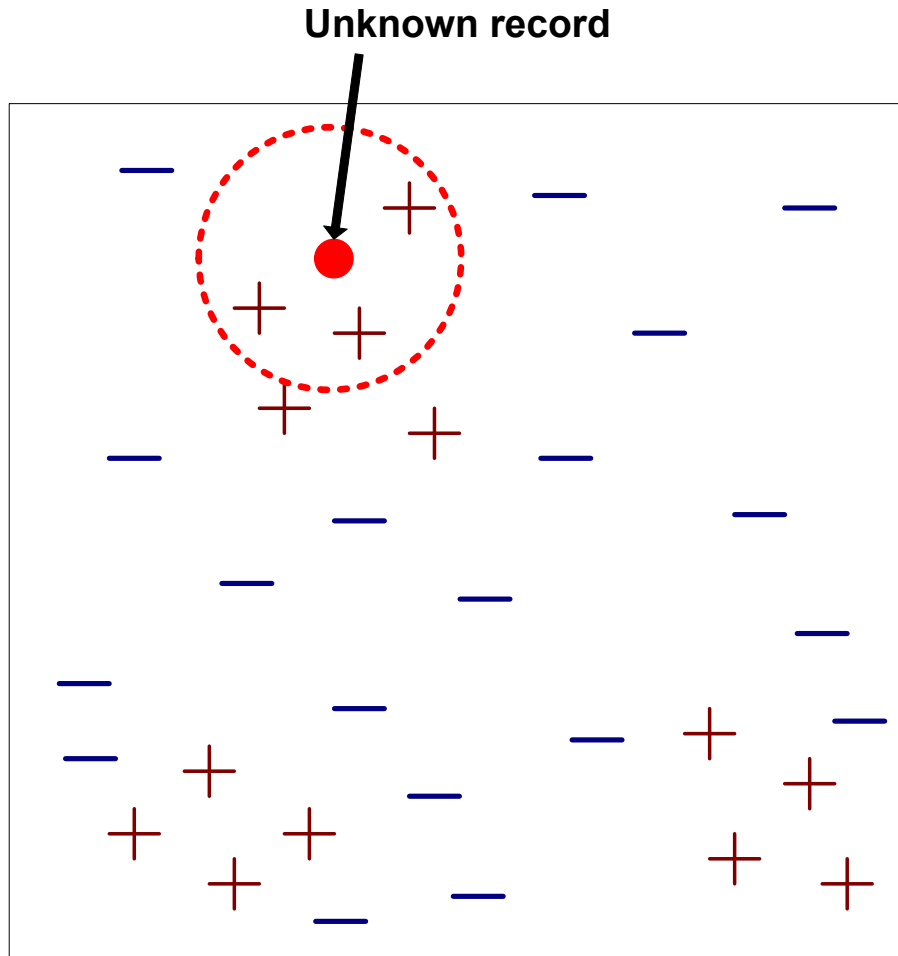
- Other metrics: Jaccard coefficient, Cosine similarity (for documents), etc.

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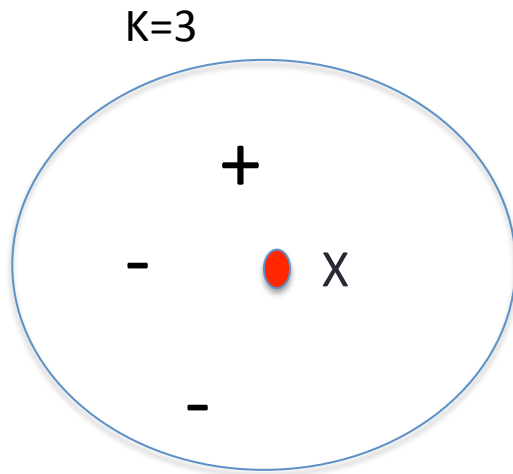
K-Nearest-Neighbor Classifiers



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Determining the class of a new instance X

- K-nearest neighbors of an instance X are data points (instances in the training set) that have the k smallest distances from X (the k most similar instances)
- What if the K-nearest neighbors have different class labels?



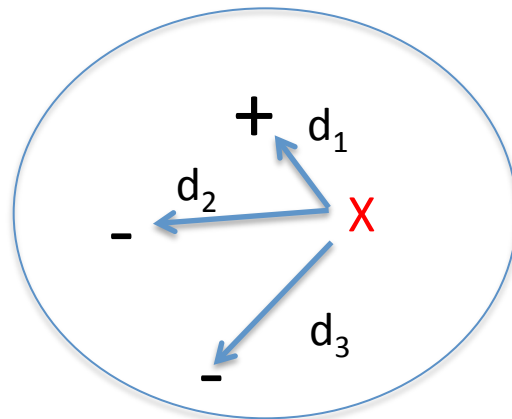
- K=3
- 1 positive and 2 negative examples
- What is the class of X?

Determining the class of a new instance X

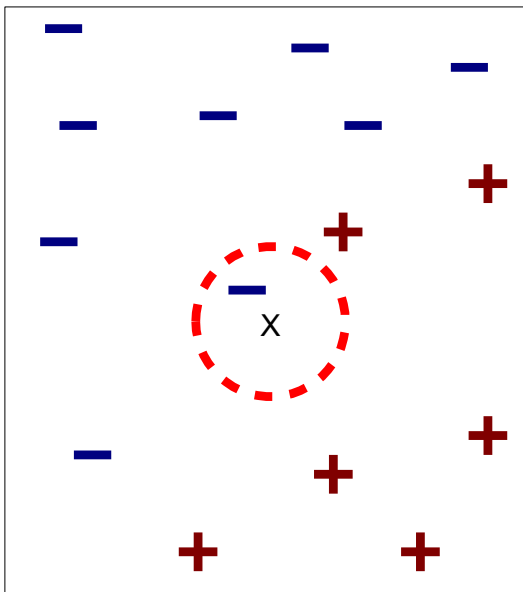
- Determining the class of a new instance X from the k nearest neighbors :
 - Each neighbor Y has associated a weight $w(Y) = 1/d^2$, where d is the distance of Y from X
 - take the majority weighted vote of class labels among the k-nearest neighbors

Nearest-Neighbor Classifiers

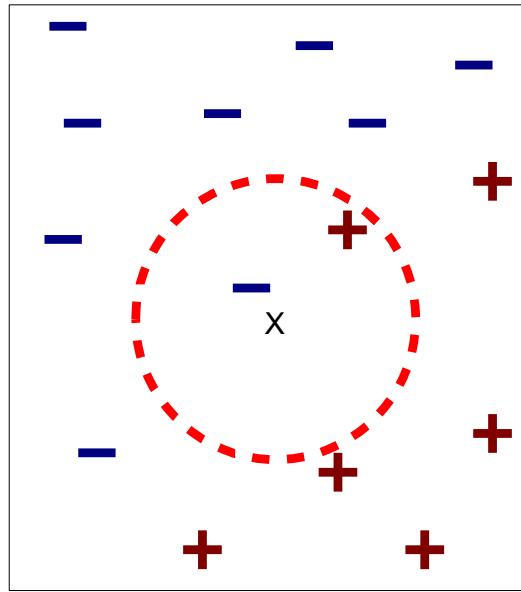
- Example: $k=3$; 1 positive example with distance $d_1=2$, and 2 negative ones, with distances $d_2=3$ and $d_3=5$.
 - $w_+ = 1/4 = 0.25$
 - $w_- = 1/9 + 1/25 = 0.15$
 - $\text{Vote} = 0.25 - 0.15 > 0$
- The new instance is classified positive



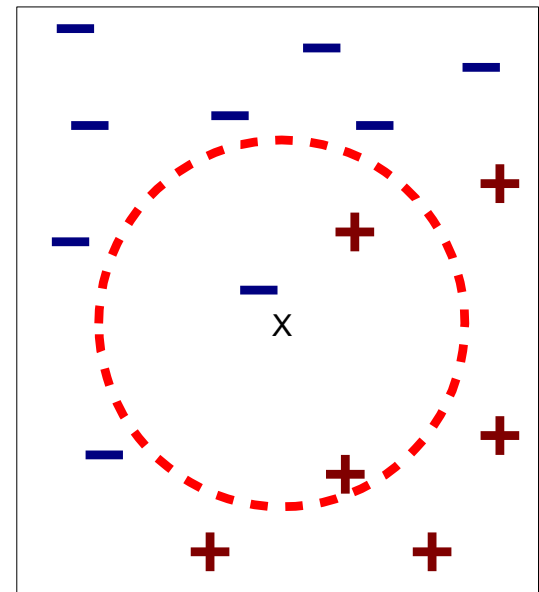
Nearest-Neighbor Classifiers



(a) 1-nearest neighbor



(b) 2-nearest neighbor

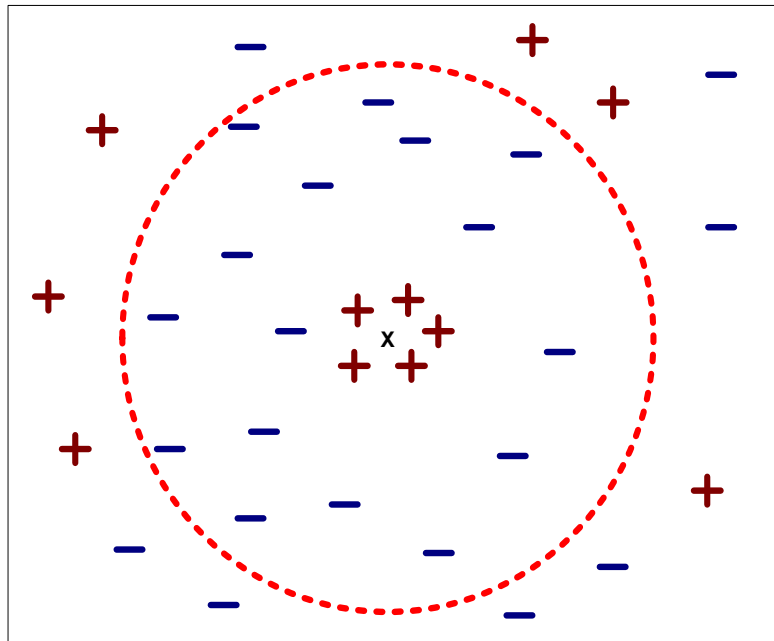


(c) 3-nearest neighbor

K-nearest neighbors of a record x are data points (instances) that have the k smallest distance to x

Nearest-Neighbor Classifiers

- 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



Conclusions

- k-NN classifiers are **lazy** learners that
 - do not build models explicitly (unlike **eager** learners such as decision tree induction and rule-based systems)
 - use a set of pre-classified instances along with similarity metrics for classifying unseen data