Lecture 5: K-nearest neighbors

Introduction to Machine Learning

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L3 MIASHS | Semestre 2

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Reminders

Reminders on previous session

Question

Can anyone remind me of the definition of supervised learning ?
Can anyone give me some kind of problems that can be solved with supervised learning ?

Principles

Main idea

K-nearest neighbors algorithm

The k-nearest neighbors algorithm is a **non-parametric supervised learning** method, which assigns to an incoming record the label issued from the plurality of votes of its k nearest neighbors.

Main idea

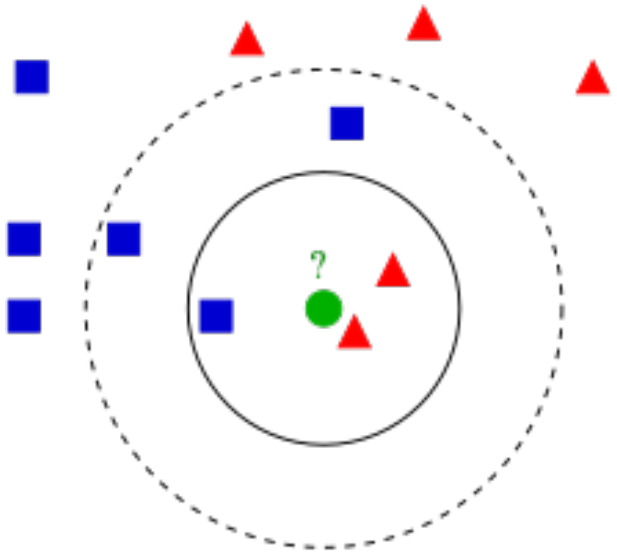
K-nearest neighbors algorithm

The k-nearest neighbors algorithm is a **non-parametric supervised learning** method, which assigns to an incoming record the label issued from the plurality of votes of its k nearest neighbors.

With an incoming data record:

- Find the $k \in \mathbb{N}$ nearest neighbors
- Assign the classification label of the most frequent labels among neighbors

Example



Can you identify a problem with certain values of k?

Example

Example: Pokemon type prediction

Training dataset:

Height	Weight	Label
45	30	Water
30	25	Water
40	35	Water
20	15	Leaf
22	18	Leaf
25	20	Leaf

Individual to classify using 1 NN and 3 NN (euclidean and manhattan distance)

Height	Weight	Label
25	31	?

Example: solution using euclidean distance

Compute distance between dataset and individual to classify:

Distance	Label
20.02	Water
7.81	Water
15.52	Water
16.76	Leaf
13.34	Leaf
11.0	Leaf

Using 1NN: *Water*
Using 3NN: *Leaf*

Hyperparameters

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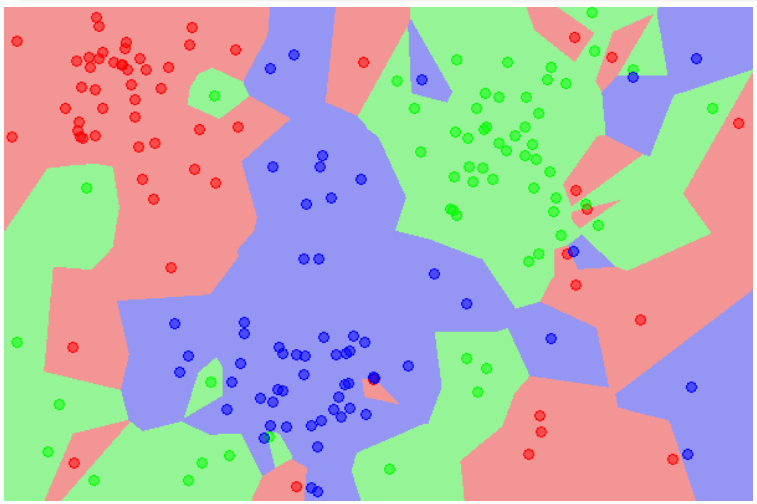

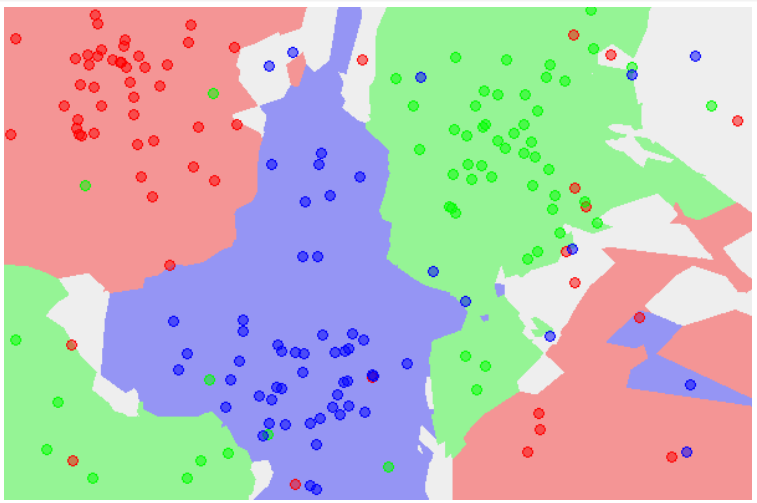

Hyperparameter selection

To select the optimum hyperparameters (distance to use, best number of neighbors), use **k-fold validation** and select the combination with the highest score (in its simplest version using a factorial design).

Advantages and limits

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- Very easy to extend to multi-class classification
- Very easy to understand
- Non-parametric algorithm (no assumption regarding data distribution)
- No previous training

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Limits:

- Very sensitive to its hyperparametrization
- Very sensitive to noise (features with little to no impact on the dataset)
- Expensive to compute
- Difficult to interprete

Questions

Questions ?