

Mesure de la qualité d'un arbre

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 - examples taken in a **reference set**
 - **degree of ease** of an example to be correctly classified

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- **Degree of quality** κ of a (fuzzy) decision tree
 - measures its ability to correctly classify difficult examples

$$\kappa(DT_j) = \frac{\sum_{i=1}^n (1 - \varepsilon(X_i)) \cdot cc_{ji}}{n}$$

with

$$cc_{ji} = \begin{cases} 1 & \text{if } X_i \text{ is correctly classified by } DT_j \\ 0 & \text{otherwise} \end{cases}$$

High Quality Trees

- Trees can be classed depending on their degree of quality
 - HQT : **high quality tree** $DT_j : \kappa(DT_j) \geq \kappa_t$
 - LQT : low quality tree

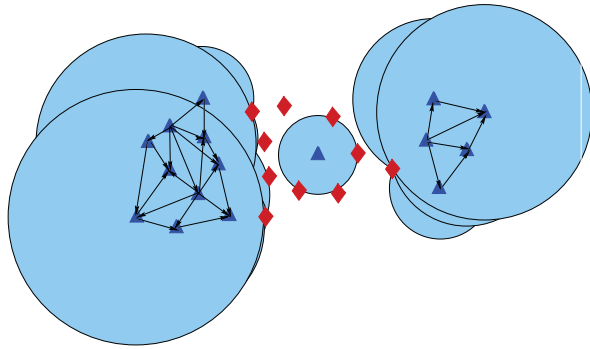
- **Objective** : to **prune** high quality trees
 - **find sub-classes** of high quality trees
 - each sub-class is represented by a single tree
 - Jason Forest's algorithm on $\{\text{HQT} \cup \text{LQT}\}$:
 - hypersphere classifier like

[Cooper 61, Reilly et al. 82, Marchand and Shawe-Taylor 02, Forest et al. 06]

- idea : to highlight subgroups in a class

The Jason Forest's Algorithm : how it works

- Considering a training set



- Sub-classes of friends are highlighted
 - here 3 distinct sub-classes are found for the blue class

The High Quality Trees Forest approach

1. Selection of examples
 - training set / reference set / test set
2. Construction of a forest of fuzzy decision trees
 - random samplings of the training set
3. Evaluation of each tree on a reference set
 - computation of the degree of ease of each example
 - computation of the degree of quality of each tree
4. Pruning of the forest
 - discovery of sub-classes among high quality trees
 - selection of the representative high quality tree of each sub-class

⇒ Size of the forest = number of sub-classes of HQT