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# Agglomerative 2-3 Hierarchical Clustering: theoretical improvements and tests

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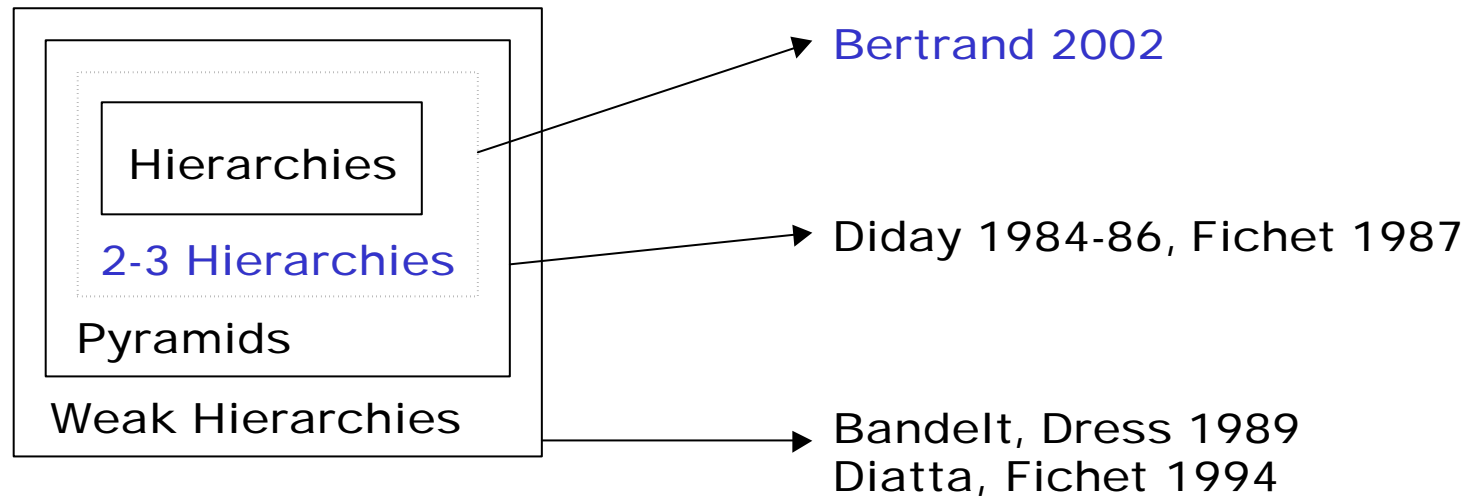
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# Outline

- The classical case of AHC
- 2-3 Hierarchies
  - Definitions
  - Properties
- Algorithm of 2-3AHC
- Analysis of complexity
- Application on simulated data
  - Experimental Validation of Complexity
  - Ongoing and Future Work

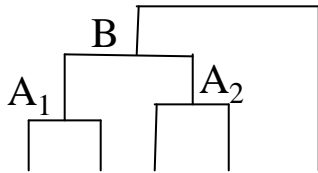
# Context



# Hierarchies (1/3)

We recall some definitions related to the hierarchical case that will be extended to the 2-3 hierarchies:

- Hierarchy:
  - each cluster is nonempty
  - $E$  and the singletons are clusters
  - each pair of clusters  $(A, B)$  is hierarchical:  
 $A \cap B \in \{E, A, B\}$



Remark: - admits at most  $n-1$  non trivial clusters

Indexed hierarchy:

- each cluster is associated to a positive real number  $f$ ,  
where  $\forall A, B \in S, A \subset B \Rightarrow f(A) < f(B)$

# Agglomerative Hierarchical Classification (2/3)

## Vocabulary:

- set inclusion order on the set of clusters:
  - predecessor/successor
  - comparable clusters
- candidate clusters (unmarked) = maximal clusters
- data input: dissimilarity  $\mathbf{d} : E \times E \rightarrow [0, \infty)$   
$$\mathbf{d}(a, b) = \mathbf{d}(b, a) > \mathbf{d}(a, a) = 0, \forall a, b \in E$$
- aggregation index (link between clusters),  $\mathbf{m}$ :
  - single linkage
  - complete linkage
  - average linkage
- usually  $f(X \hat{=} Y) = \mathbf{m}(X, Y)$

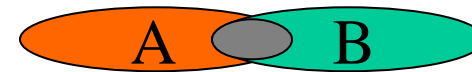
## Algorithm AHC (3/3)

1. Initialisation:  $iter \leftarrow 0$ ; Clusters are the singletons of set  $E$ .  
 $f \leftarrow 0$ ;
2.  $iter \leftarrow iter + 1$ ;  
Merge  $X$  and  $Y$  which are - in the sense of  $m$  - the two nearest clusters; compute  $f(X \dot{\cup} Y)$
3. Reduction: Eliminate the successors found on the same level  $f$  with their predecessor, if there are any
4. Update  $m$ , predecessor links, successor links
5. Stopping rule: Repeat step 2-4, until the set  $E$  becomes a cluster

## 2-3 Hierarchies: Definitions

- Proper intersection:

- *A properly intersects B, if  $A \cap B \neq \{A, B\}$*



Concept: - in a 2-3 hierarchy, for any three clusters at least two pairs of them are hierarchical

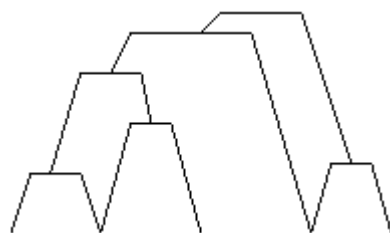
- 2-3 Hierarchy [Bertrand 2002]:

- each cluster is nonempty
- $E$  and singletons are clusters
- the proper intersection of two clusters is also a cluster
- each cluster properly intersects no more than one other cluster

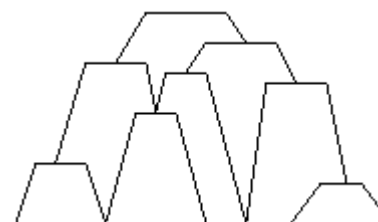
## 2-3 Hierarchies: Properties

[Bertrand 2002]

- The number of elements of a 2-3 hierarchy that are not reduced to singletons, is at most  $\left\lceil \frac{3}{2}(n-1) \right\rceil$
- Each 2-3 hierarchical set system on  $E$  is a collection of intervals of some linear order defined on  $E$ .



2-3 Hierarchy



Pyramid

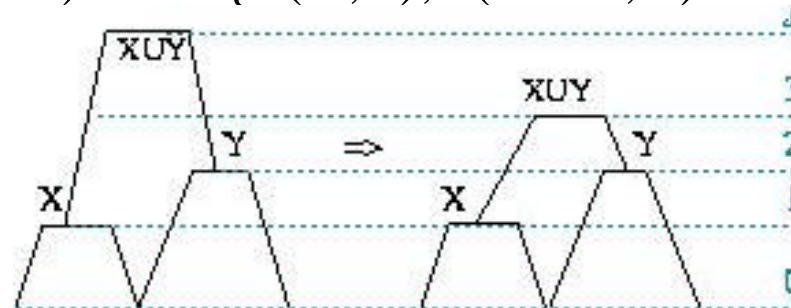


# Algorithm of 2-3AHC

1. Initialisation:  $iter \leftarrow 0$ ; Clusters are the singletons of set  $E$ .  
 $f \leftarrow 0$ ;
2.  $iter \leftarrow iter + 1$ ;  
Merge  $X$  and  $Y$  which are - in the sense of  $m$  - the two nearest *non-comparable* clusters, such that at least one of them is maximal; compute  $f(X \dot{\cup} Y)$
3. Merge  $X \dot{\cup} Y$  and the other predecessor of  $X$  or  $Y$ , if it exists.  
compute  $f(X \dot{\cup} Y)$
4. Reduction: Eliminate the successors found on the same level  $f$  with their predecessor, if there are any
5. Update  $m$ , predecessor links, successor links
6. Stopping rule: Repeat step 2-5, until the set  $E$  becomes a cluster

# Algorithm of 2-3AHC

- Generalizes the AHC:
  - a cluster can be merged with two different clusters
- Double single linkage [Jullien, Bertrand 2002]:  
 $f(X \cup Y) = \text{Min}\{\mathbf{m}(X, Y), \mathbf{m}(X \cup Y, Z) : Z \text{ candidate cluster}\}$



- Complexity:  $O(n^2 \log n)$

# Analysis of Complexity (1/3)

We use an ordered dissimilarity matrix on three levels:

- dissimilarity values
- cardinality of the two clusters
- lexicographical order

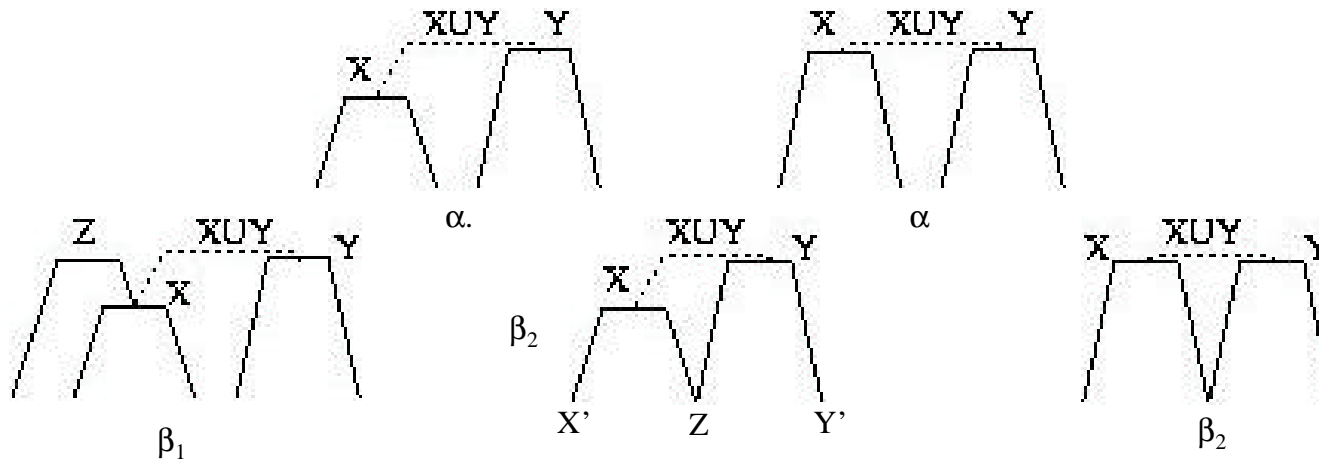
Step 1. Initialisation: Compute and order the dissimilarity matrix,  $O(n^2 \log n)$

Step 2. Merge X and Y ... : Retrieve (X,Y) from the data structure, and create  $X \dot{\cup} Y$ ,  $O(1)$

Step 3. Merge  $X \dot{\cup} Y$  and ... : Intermediate merging with  $O(n)$  complexity

# Analysis of Complexity (2/3)

Step 4. Reduction: We have five possible cases of reduction when merging a cluster:



- eliminate the successors found on the same level with their predecessor
- complexity  $O(n)$

## Analysis of Complexity (3/3)

Step 5. Update  $m$  :

- compute new dissimilarities and store them in the matrix,  $O(n \log n)$
- eliminate dissimilarities containing non candidates clusters,  $O(n \log n)$

Total complexity of the algorithm:

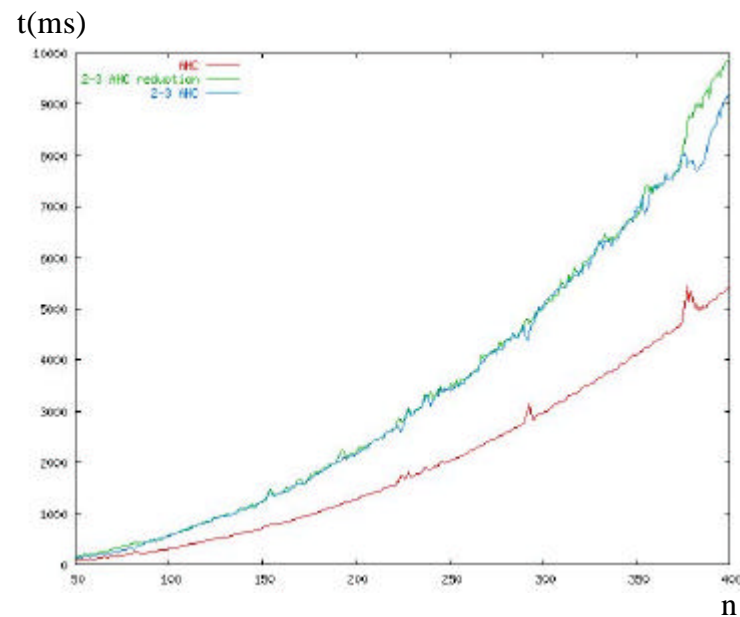
$$O(n^2 \log n) + n \times O(n \log n) \rightarrow O(n^2 \log n)$$

step 1.            steps 2. - 5.

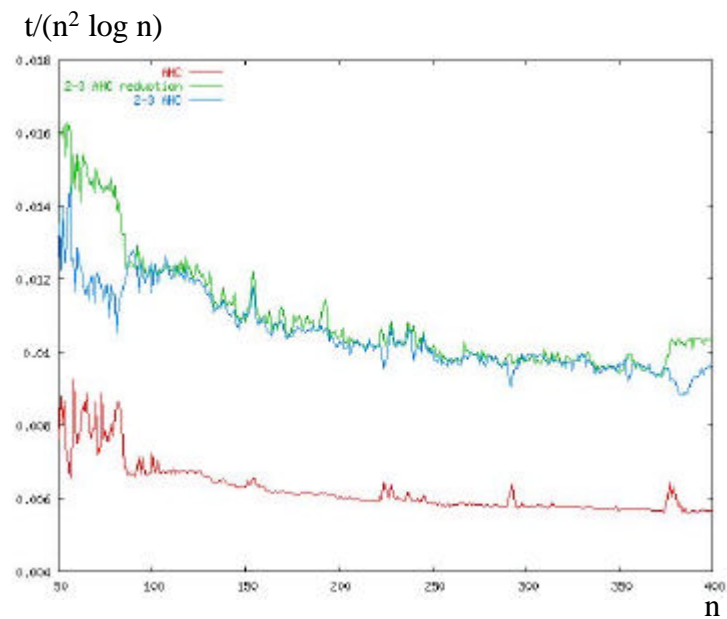
# Application

## Simulated data:

- complexity  $\Rightarrow O(n^2 \log n)$



Execution times



Complexity

# Conclusions

## Contributions:

- a new formulation of the 2-3 AHC algorithm
- a reduction of complexity
- a first implementation of the 2-3 AHC algorithm (Java) and its integration in CBR\*Tools, a Case Based Reasoning framework
- an experimental validation of the complexity on simulated data

## Ongoing and Future work:

- study of the quality of the 2-3 AHC compared with AHC and other classification methods
- study of the applicability of 2-3 AHC in the context of Web Usage Mining