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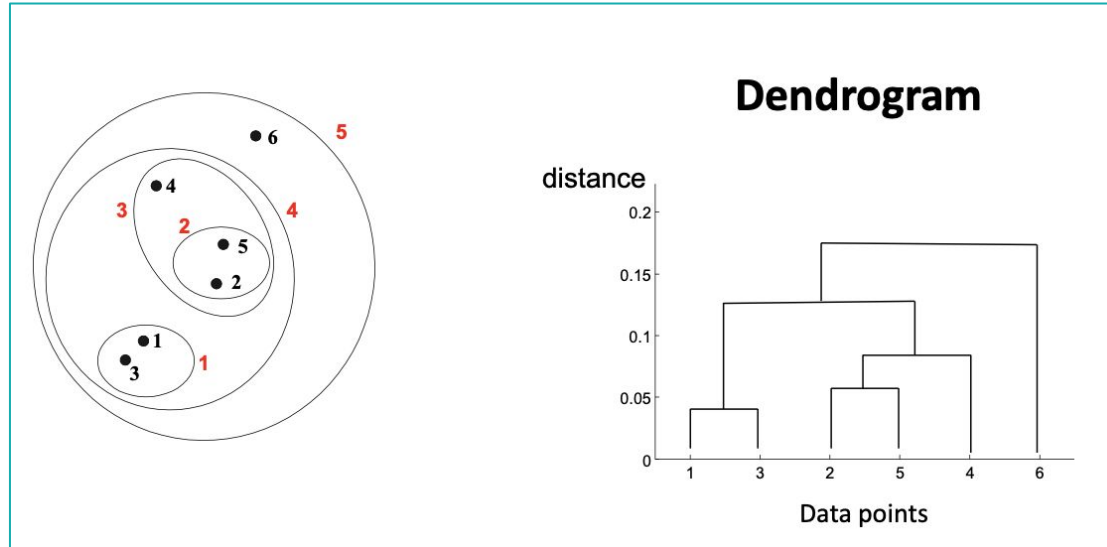
**Hierarchical Clustering**



# Hierarchical Clustering

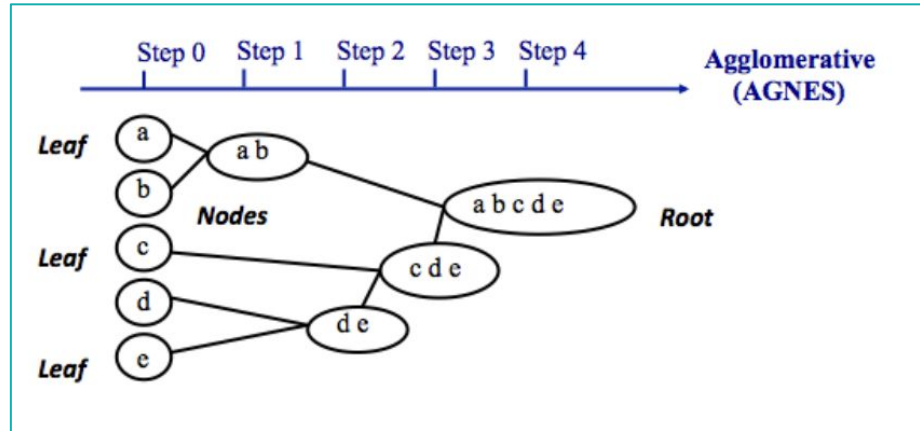
Produces a set of nested clusters organized as a hierarchical tree called a dendrogram.

The dendrogram shows at what distance points join into a cluster.



# Agglomerative Clustering - Overview

Start with the points as individual clusters - At each step, merge the closest pair of clusters until only one cluster (or k clusters) remaining

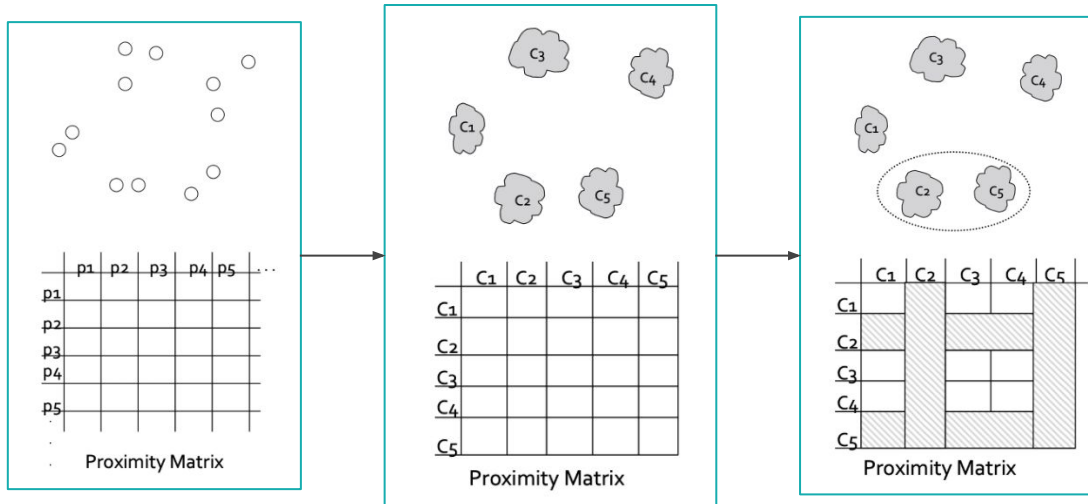


Key operation is the computation of the proximity of two clusters.



# Agglomerative Clustering Algorithm

1. Start with clusters of individual points and a proximity matrix
2. After some merging steps, we have some clusters
3. We want to merge the two closest clusters (C2 and C5) and update the proximity matrix.



Compute the **proximity matrix**

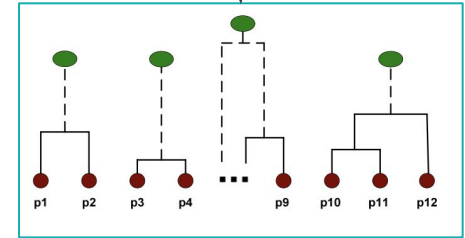
Let each data point be a cluster

**Repeat**

**Merge** the two closest clusters

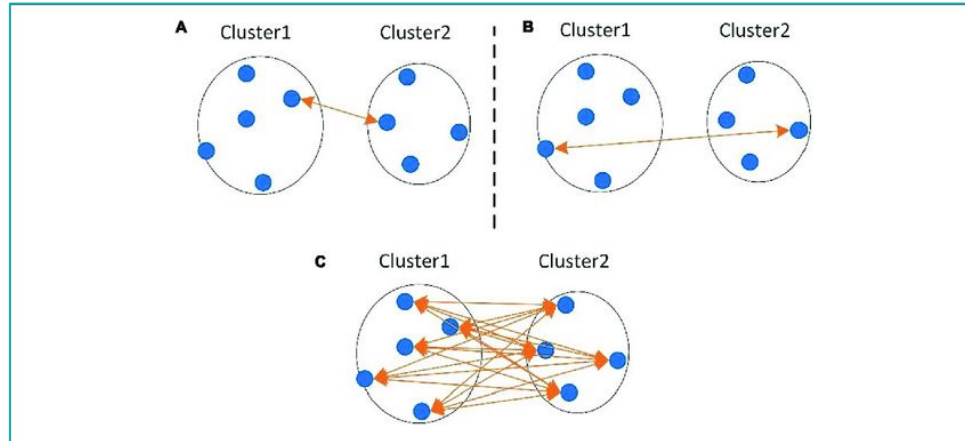
**Update** the proximity matrix

**Until** only a single cluster remains



# Measuring the distance of two clusters

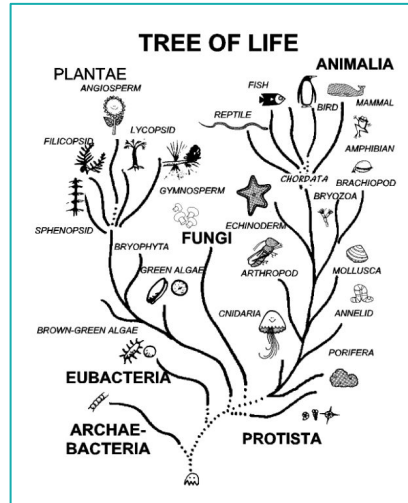
1. Single link method: Distance between two closest data points in the two clusters.
2. Complete link method: Distance of two furthest data points in the two clusters.
3. Average link: Average distance of all pairwise distances between the data points in two clusters.



# Strengths of Hierarchical Clustering

We do not have to assume any particular number of clusters:

- Any desired number of clusters can be obtained by 'cutting' the dendrogram at the proper level.
  - They may correspond to meaningful taxonomies: Example in biological sciences.



# Limitations of Hierarchical Clustering

- **Greedy**: Once a decision is made to combine two clusters, it cannot be undone
- **No global objective function** is directly minimized
- Sensitivity to **noise** and **outliers**
- Difficulty handling different sized clusters and convex shapes
- Chaining, breaking large clusters



**Much obliged.**

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