



# Hierarchical Data Analysis

## What is Hierarchical Clustering?

# Objective

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Objective

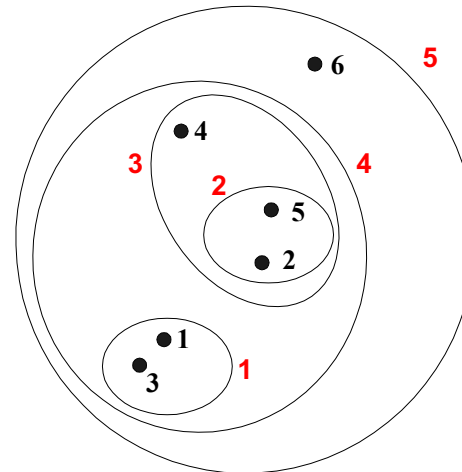
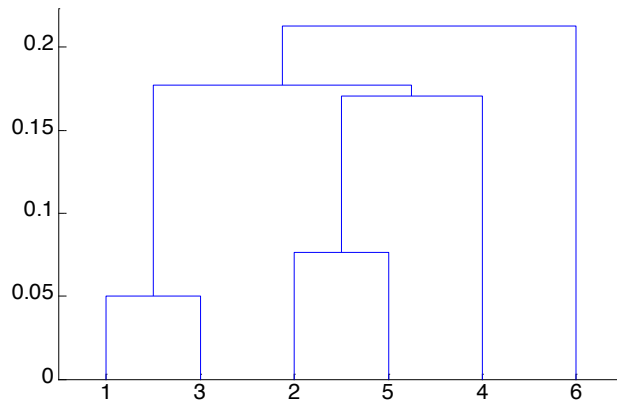
Apply methods of  
hierarchical data analysis

# Hierarchical Clustering

| Produces a set of **nested clusters** organized as a hierarchical tree

| Can be visualized as a **dendrogram** (along with other options)

- A tree-like diagram that records the sequences of merges or splits



# Strengths of Hierarchical Clustering



| No assumptions on the number of clusters

- Any desired number of clusters can be obtained by ‘cutting’ the dendrogram at the proper level

| Hierarchical clusterings may correspond to meaningful taxonomies

- Example in biological sciences (e.g., phylogeny reconstruction, etc), web (e.g., product catalogs) etc

# Hierarchical Clustering

## | Agglomerative:

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

## | Divisive:

- Start with one, all-inclusive cluster
- At each step, split a cluster until each cluster contains a point (or there are  $k$  clusters)

| Traditional hierarchical algorithms use a similarity or distance matrix

- Merge or split one cluster at a time

# Complexity of Hierarchical Clustering

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- | Distance matrix is used for deciding which clusters to merge/split

- | At least quadratic in the number of data points

- | Not usable for large datasets