

# Data Mining

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Department of CS - DM - UHD

2



## Road map

- Cluster Analysis: Basic Concepts
- Partitioning Methods
- Hierarchical Methods
- What is Hierarchical Clustering
- General Steps Of Hierarchical Clustering
- Methods of Hierarchical Clustering
- Agglomerative (bottom up)
- Divisive (top down)
- Dendrogram
- Summary

### Cluster Analysis: Basic Concepts and Methods

- Cluster Analysis: Basic Concepts
- Partitioning Methods
- Hierarchical Methods —
- Evaluation of Clustering
- Summary

## What is Hierarchical Clustering

- In data mining and statistics, hierarchical clustering (also called hierarchical cluster analysis or HCA) is a method of cluster analysis which seeks to build a hierarchy of clusters.
- The idea is to build a binary tree of the data that successively merges similar groups of points

### General Steps Of Hierarchical Clustering

# Given a set of N items to be clustered, the basic process of hierarchical clustering is this:

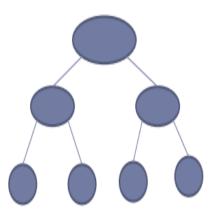
- 1. Start by assigning each item to a cluster, so that if you have N items, you now have N clusters, each containing just one item.
- 2. Find the closest (most similar) pair of clusters and merge them into a single cluster, so that now you have one cluster less.
- 3. Compute similarities between the new cluster and each of the old clusters.
- 4. Repeat steps 2 and 3 until all items are clustered into K number of clusters

### **Methods of Hierarchical Clustering**

There are two main types of hierarchical clustering: group data objects into a tree of clusters

#### Hierarchical methods can be

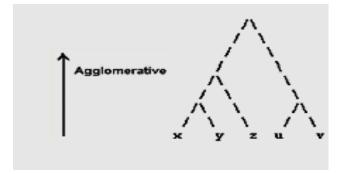
- \* Agglomerative: bottom-up approach
- **❖ Divisive**: top-down approach



# Hierarchical Clustering Agglomerative and Divisive Methods

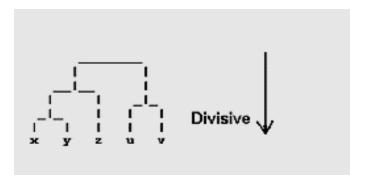
#### **Agglomerative (bottom up)**

- 1. Start with 1 point (singleton)
- 2. Recursively add two or more appropriate clusters
- 3. Stop when k number of clusters is achieved.



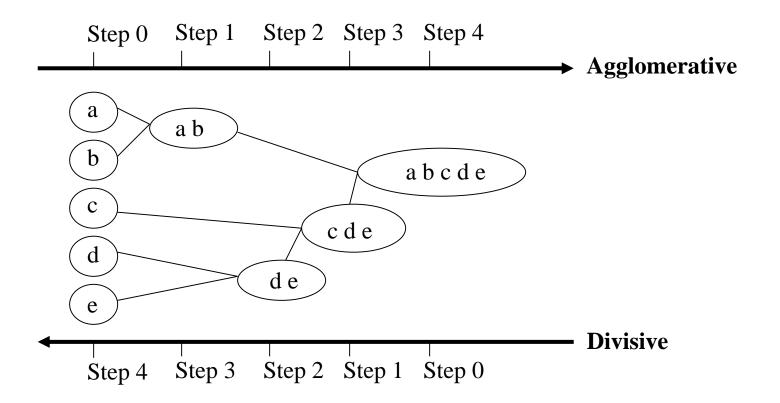
#### **Divisive (top down)**

- 1. Start with a big cluster
- 2. Recursively divide into smaller clusters
- 3. Stop when k number of clusters is achieved.

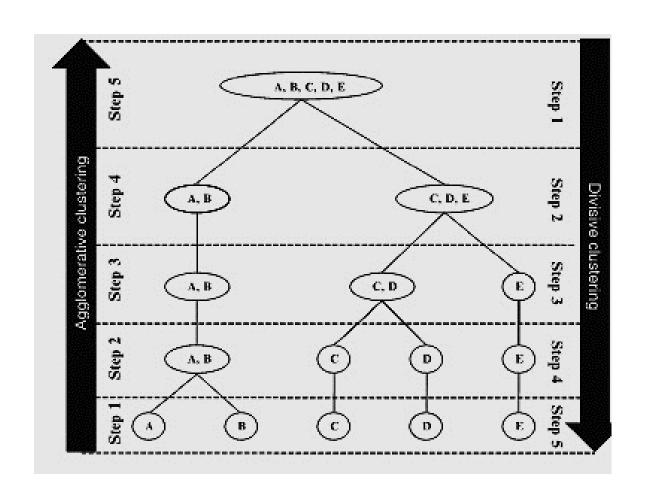


# Hierarchical Clustering Agglomerative and Divisive Methods

Use similarity matrix as clustering criteria. Stop when k number of clusters is achieved.



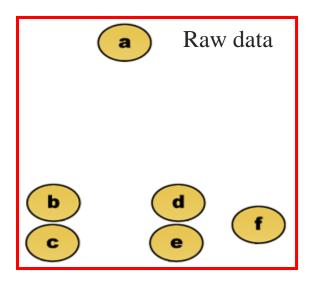
# Hierarchical Clustering Agglomerative and Divisive Methods

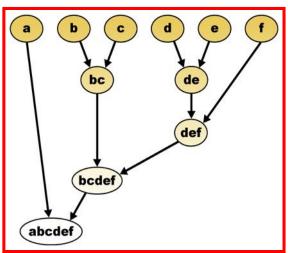


### Hierarchical Agglomerative General Algorithm

In this example, after the **second step** of the agglomerative algorithm will yield clusters:- {**a**} {**b c**} {**d e**} {**f**}.

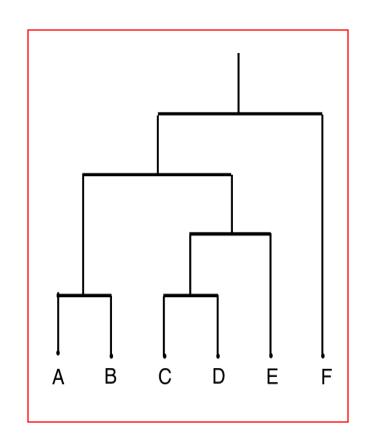
In the third step will yield clusters {a} {b c} {d e f}, which is a clustering, in the fourth step will give a small number but larger clusters that are {a} {b c d e f} and finally will yield cluster of {a b c d e f}





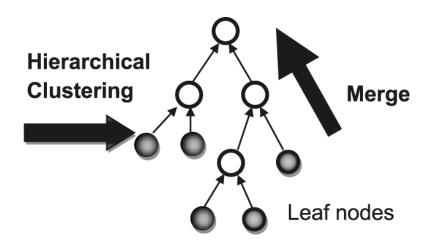
## Dendrogram Hierarchical Clustering

- Dendrogram: a tree data structure which illustrates hierarchical clustering techniques.
- Each level shows clusters for that level.
  - Leaf individual clusters
  - Root one cluster

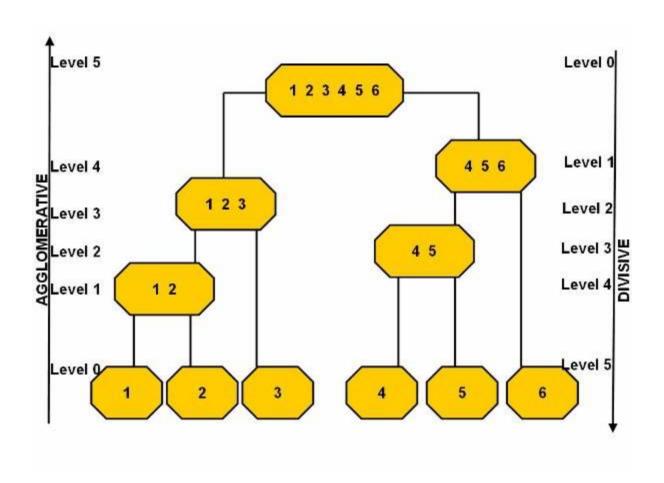


### Dendrogram: Shows How Clusters are Merged

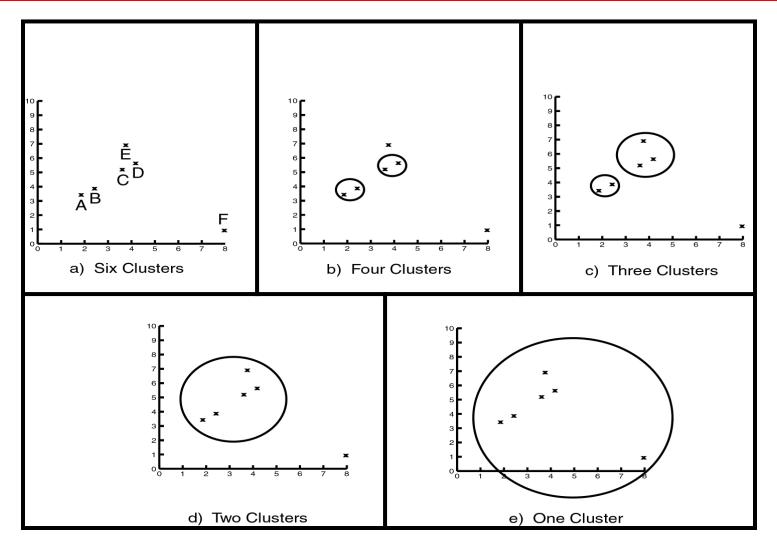
- Show how to merge clusters hierarchically
- Decompose data objects into a multi-level of a tree of clusters
- A clustering of the data objects: giving the dendrogram at the desired level
  - Each connected component forms a cluster



# Levels of Clustering



# Levels of Traditional Clustering

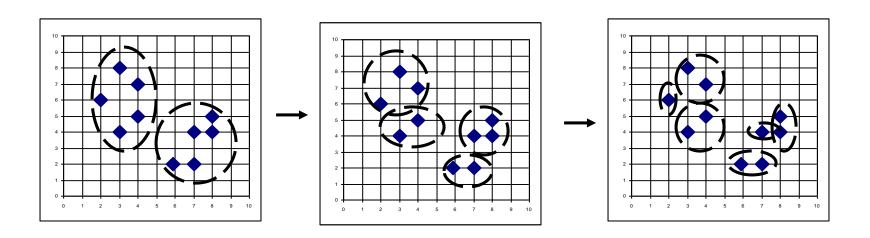


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# Traditional Hierarchical Clustering Divisive Method

#### **Divisive (top down)**

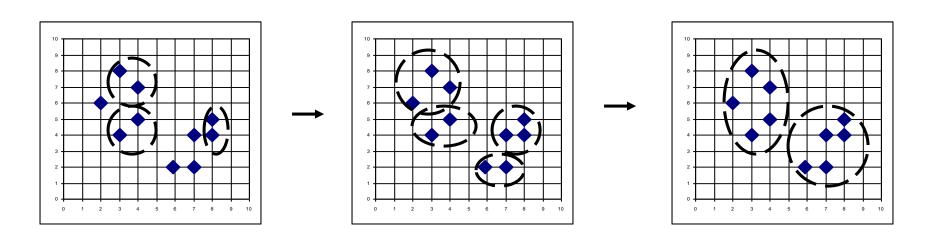
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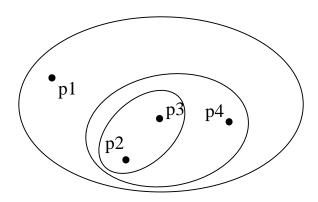


# Traditional Hierarchical Clustering Agglomerative Method

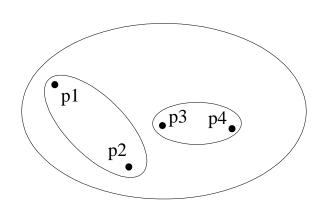
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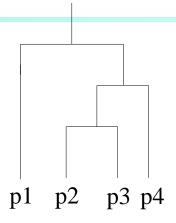




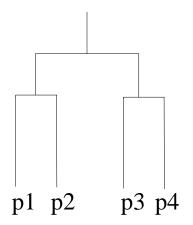
**Traditional Hierarchical Clustering** 



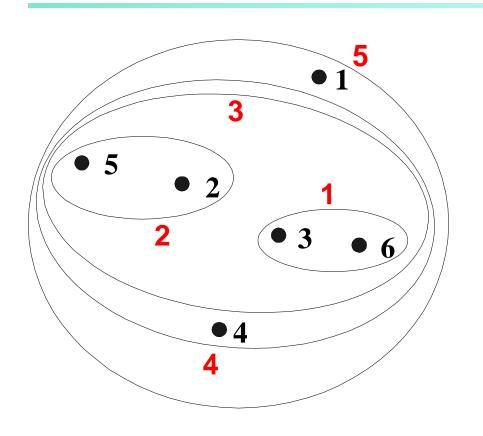
**Non-traditional Hierarchical Clustering** 

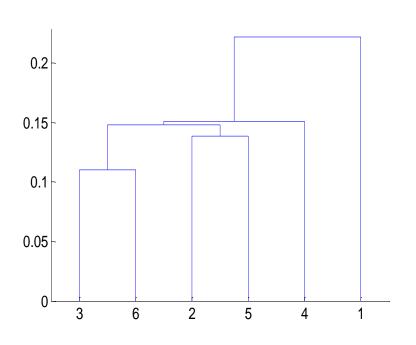


**Traditional Dendrogram** 



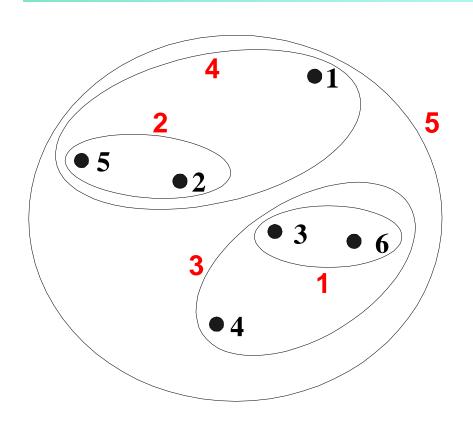
**Non-traditional Dendrogram** 

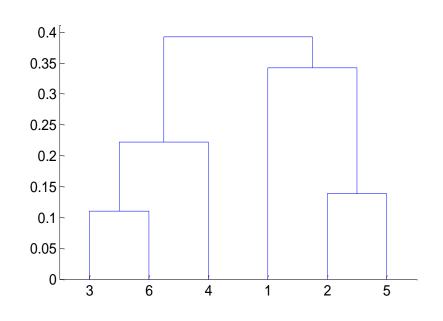




**Nested Clusters** 

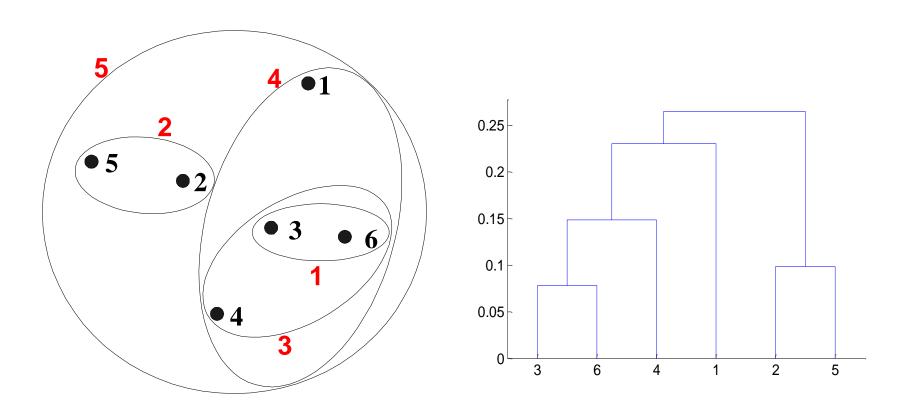
**Dendrogram** 





**Nested Clusters** 

**Dendrogram** 



**Nested Clusters** 

**Dendrogram** 

