# Hierarchical Data Analysis Agglomerative Clustering



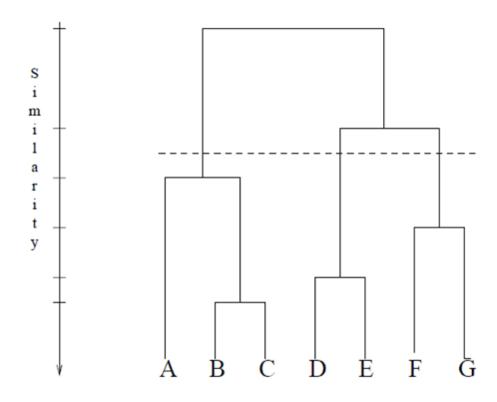
## **Objective**



Apply methods of hierarchical data analysis

## **Agglomerative Clustering Algorithm**

Most popular hierarchical clustering technique



## **Agglomerative Clustering Algorithm**

#### Basic algorithm

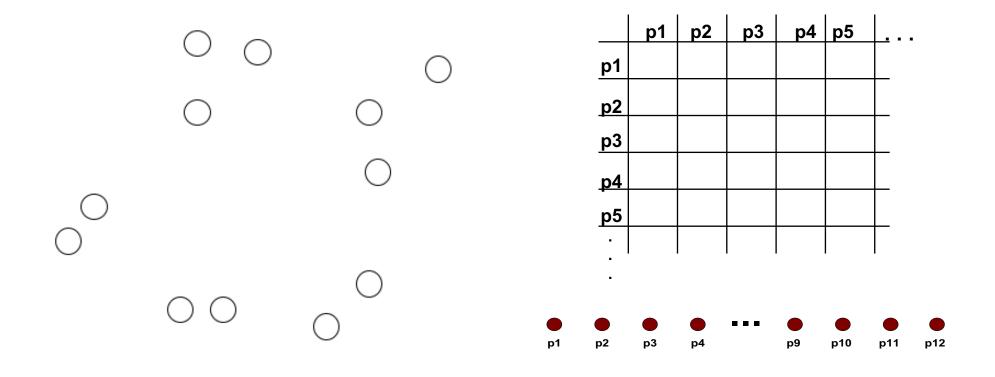
- 1. Compute the distance matrix between the input data points
- 2. Let each data point be a cluster
- 3. Repeat
- 4. Merge the two closest clusters
- 5. Update the distance matrix
- 6. Until only a single cluster remains

## **Agglomerative Clustering Algorithm**

- Key operation is the computation of distance between two clusters
- Different definitions of the distance between clusters lead to different algorithms

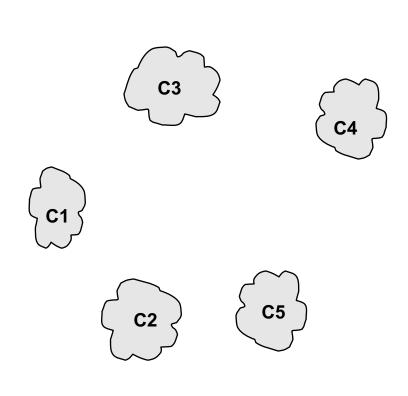
## Hierarchical Clustering: Input/Initial Setting

Start with clusters of individual points and a distance/proximity matrix



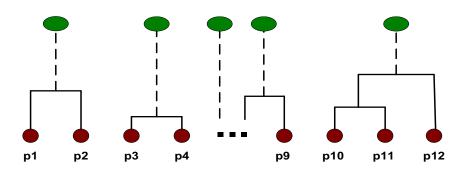
### **Intermediate State**

After some merging steps, we have some clusters



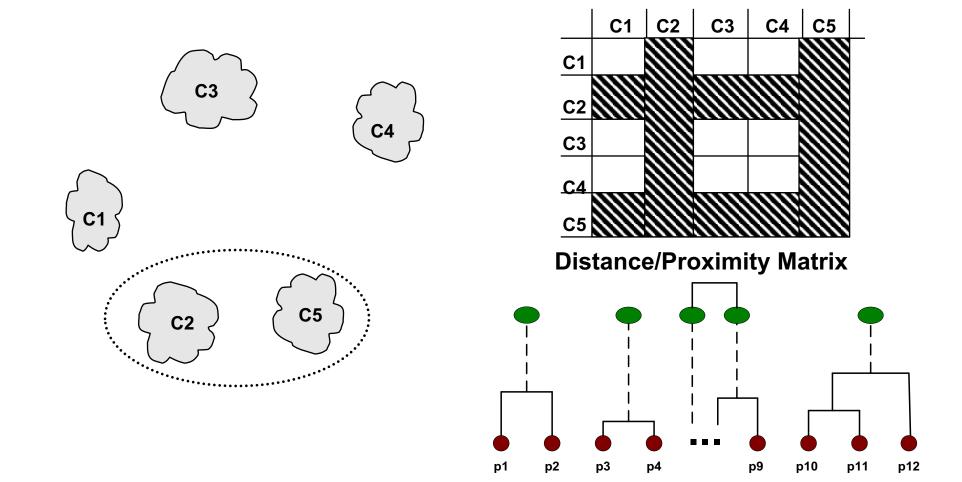
	<b>C1</b>	C2	<b>C</b> 3	C4	<b>C</b> 5
<u>C1</u>					
<b>C2</b>					
<b>C</b> 3					
C4					
<b>C</b> 5					

#### **Distance/Proximity Matrix**



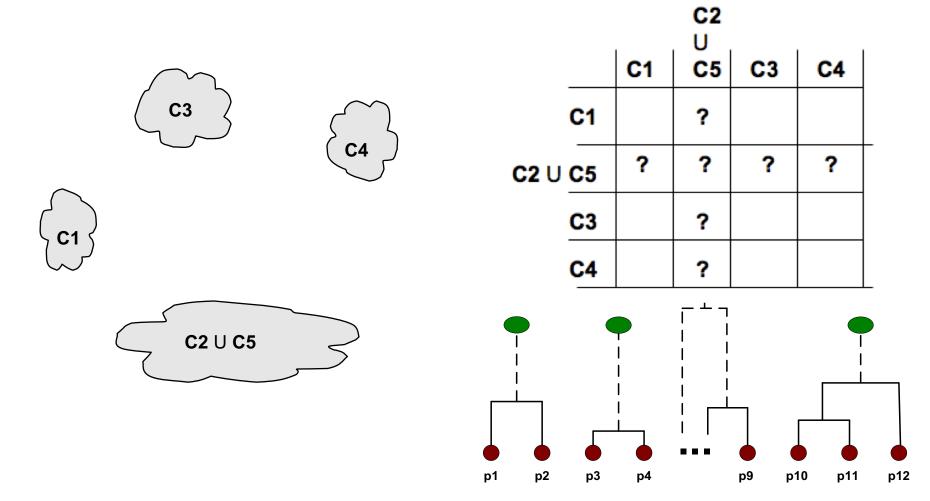
#### **Intermediate State**

Merge two closest clusters (C2 and C5) and update distance matrix



## **After Merging**

"How do we update the distance matrix?"



#### Distance between two clusters

Each cluster is a set of points

- How do we define distance between two sets of points?
- Lots of alternatives
- Not an easy task

#### Distance between two clusters

Single-link distance between clusters  $C_i$  and  $C_j$  is the *minimum distance* between any object in  $C_i$  and any object in  $C_j$ 

The distance is defined by the two most similar objects

$$D_{sl}(C_i, C_j) = \min_{x,y} \left\{ d(x, y) \middle| x \in C_i, y \in C_j \right\}$$