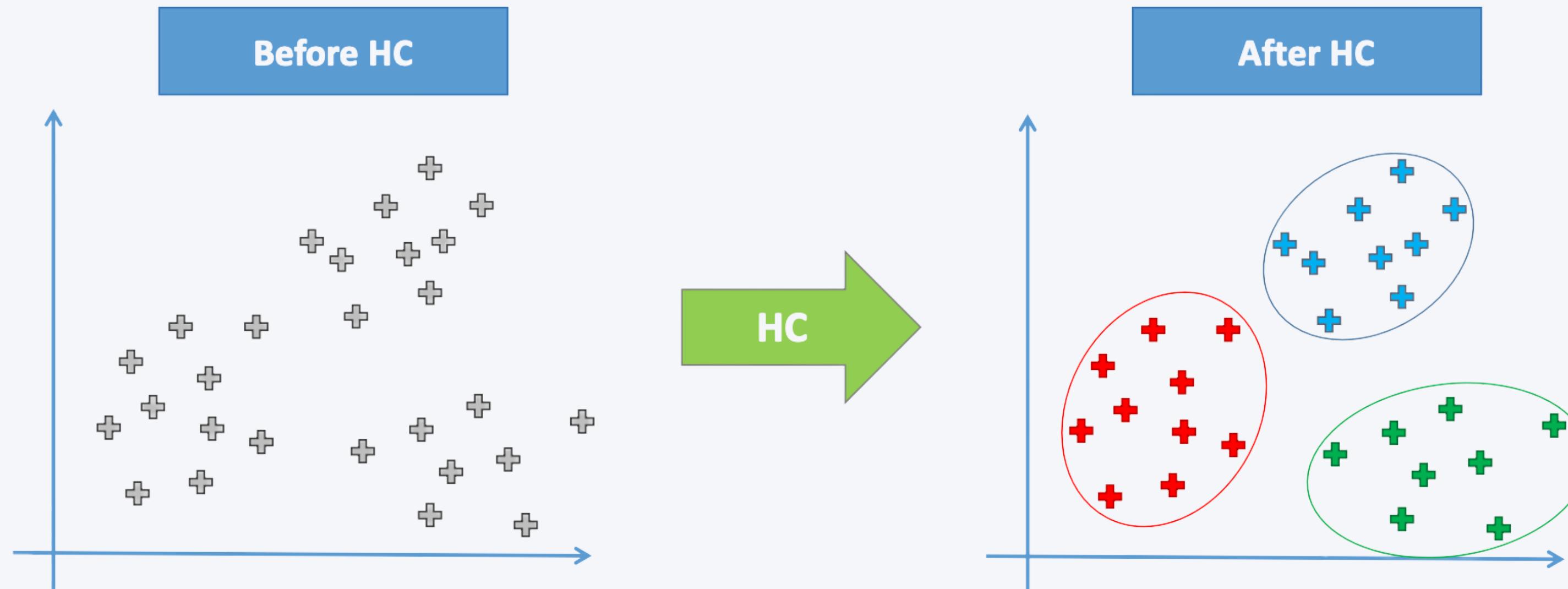


# Hierarchical Clustering

# Hierarchical Clustering

## What HC does ?



Same as K-Means but different process

# Hierarchical Clustering

Two types of HC:

**Agglomerative & Divisive**

Bottom-Up Approach

# Hierarchical Clustering

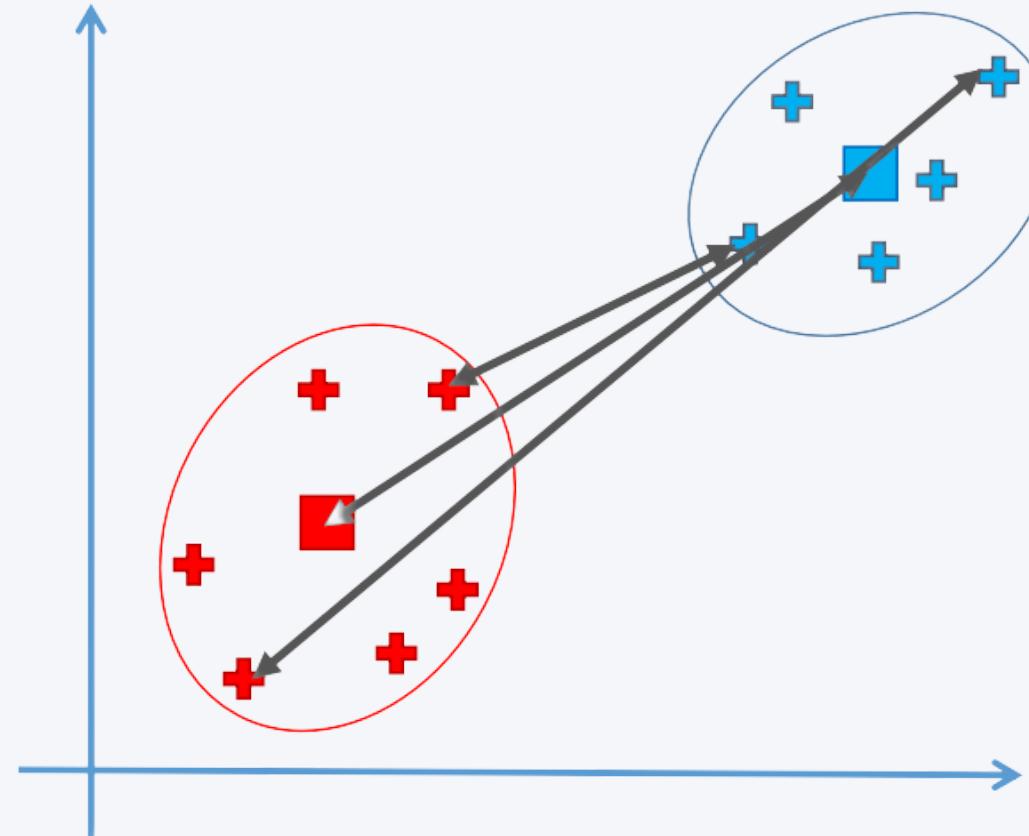
## Agglomerative HC

- **STEP 1:** Make each data point a single-point cluster -> That forms N clusters
- **STEP 2:** Take the two closest data points and make them **one cluster** -> That forms N-1 clusters
- **STEP 3:** Take the two **closest clusters** and make them one cluster -> That forms N - 2 clusters
- **STEP 4:** Repeat STEP 3 until there is **only one cluster**



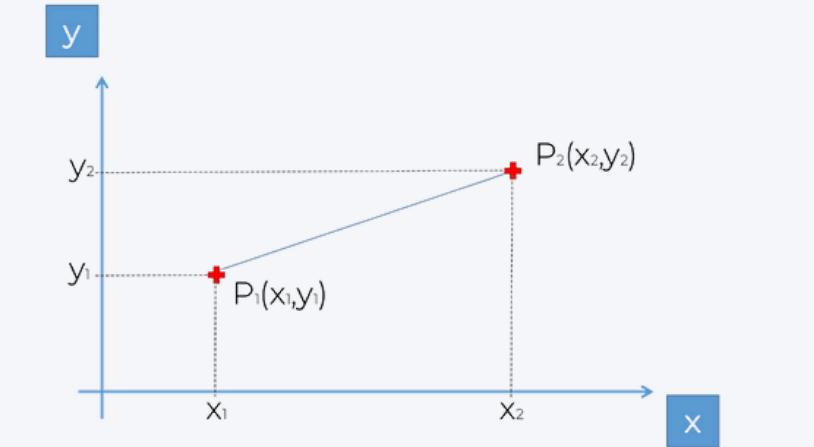
# Hierarchical Clustering

## Agglomerative HC



Distance Between Two Clusters:

- Option 1: Closest Points
- Option 2: Furthest Points
- Option 3: Average Distance
- Option 4: Distance Between Centroids



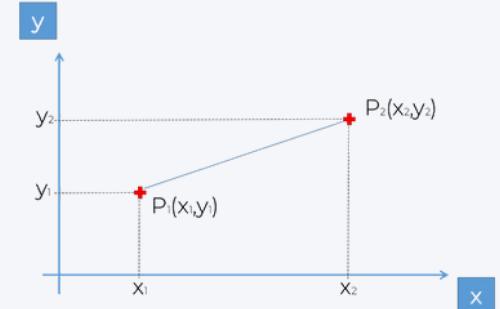
$$\text{Euclidean Distance between } P_1 \text{ and } P_2 = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

- Euclidean Distance is basically more natural

- **What is actually Distance between clusters ?** shortest ? largest ? average ? or distance between centroids ?
- You need to find that based on business usecase !! choosing distance is **important** !!

# Hierarchical Clustering

## Agglomerative HC



$$\text{Euclidean Distance between } P_1 \text{ and } P_2 = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

Consider the following dataset of  $N = 6$  data points



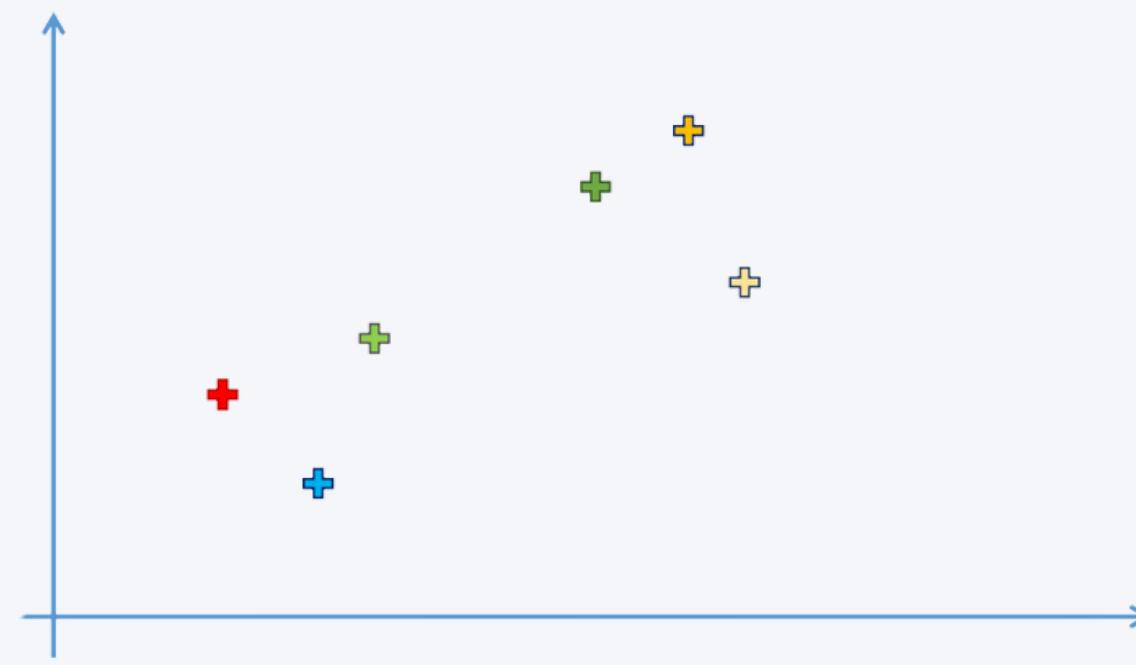
STEP 1: Make each data point a single-point cluster → That forms 6 clusters



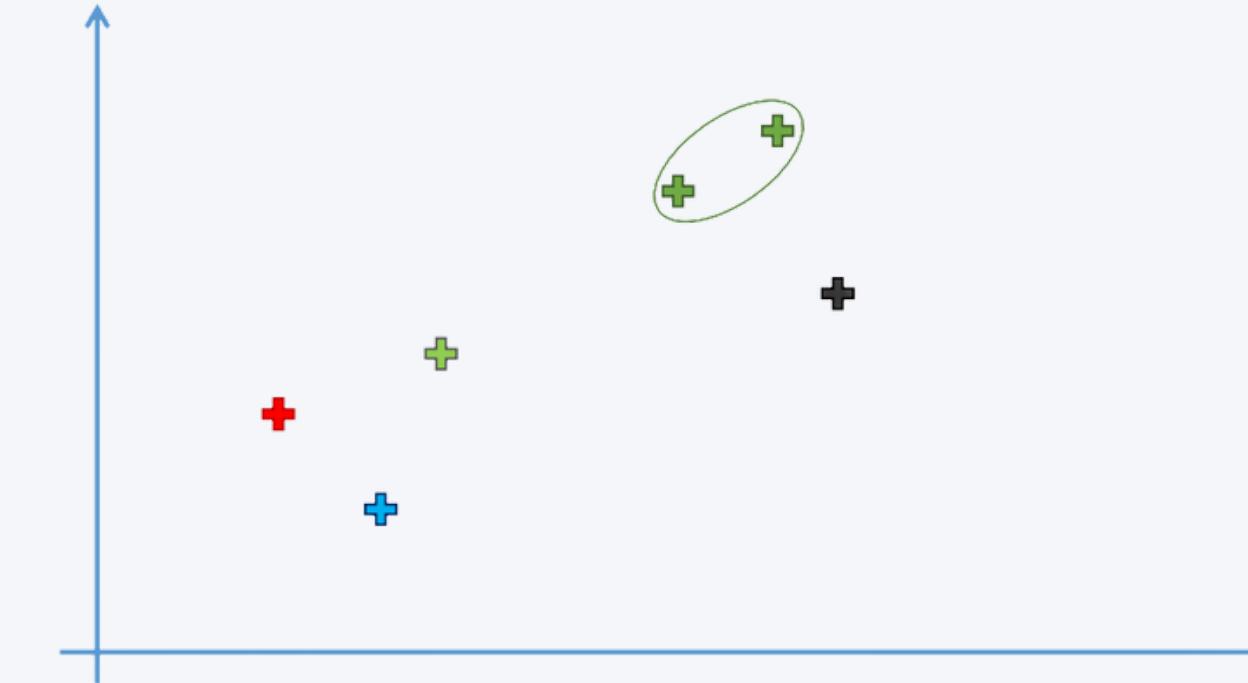
# Hierarchical Clustering

## Agglomerative HC

STEP 1: Make each data point a single-point cluster → That forms 6 clusters



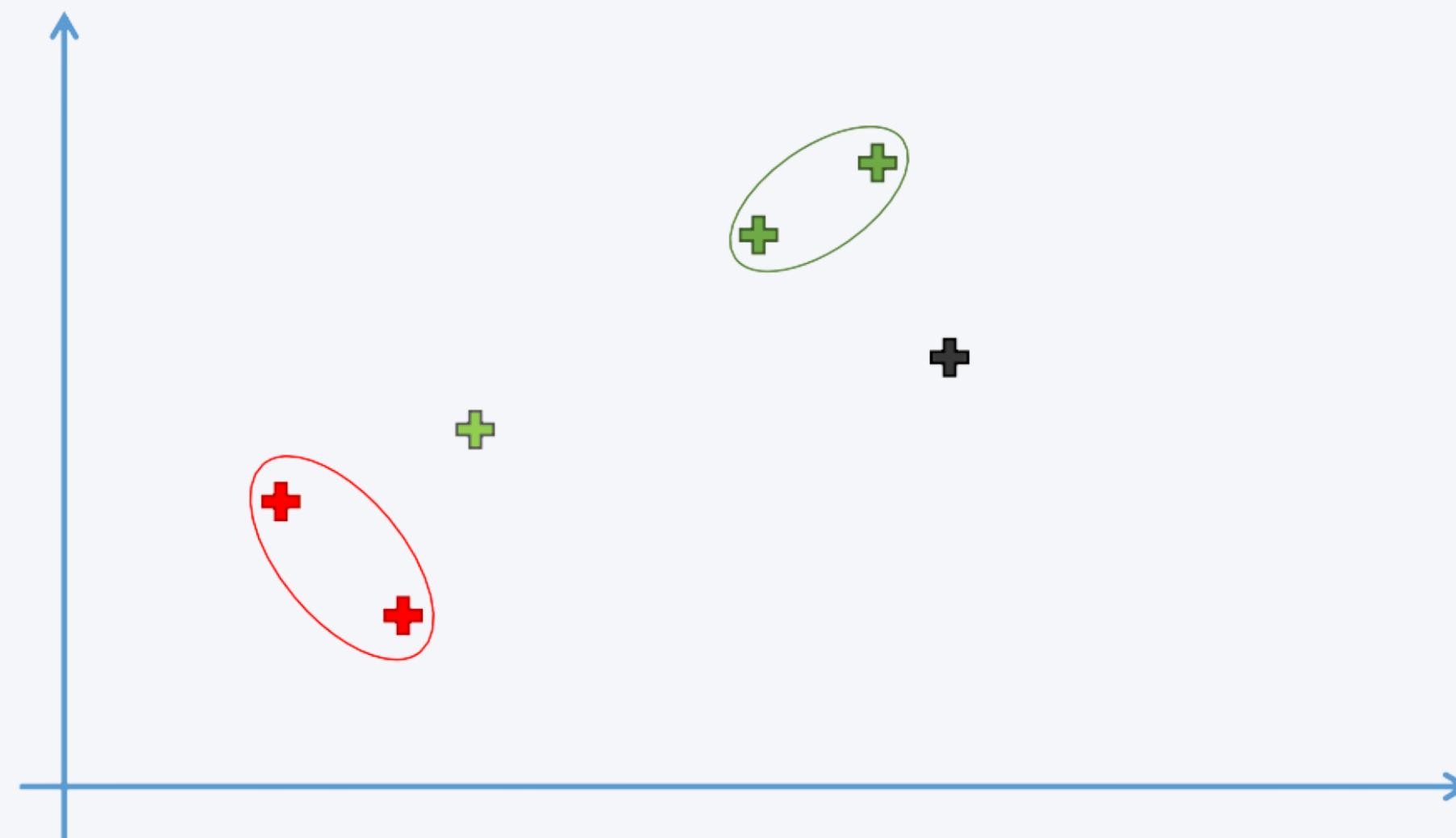
STEP 2: Take the two closest data points and make them one cluster  
→ That forms 5 clusters



# Hierarchical Clustering

## Agglomerative HC

STEP 3: Take the two closest clusters and make them one cluster  
→ That forms 4 clusters



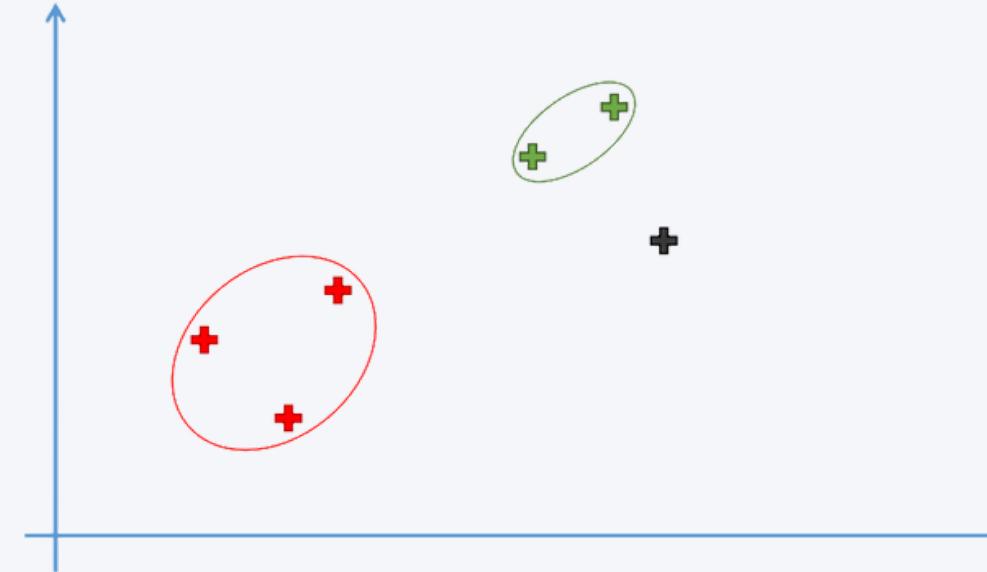
# Hierarchical Clustering

## Agglomerative HC

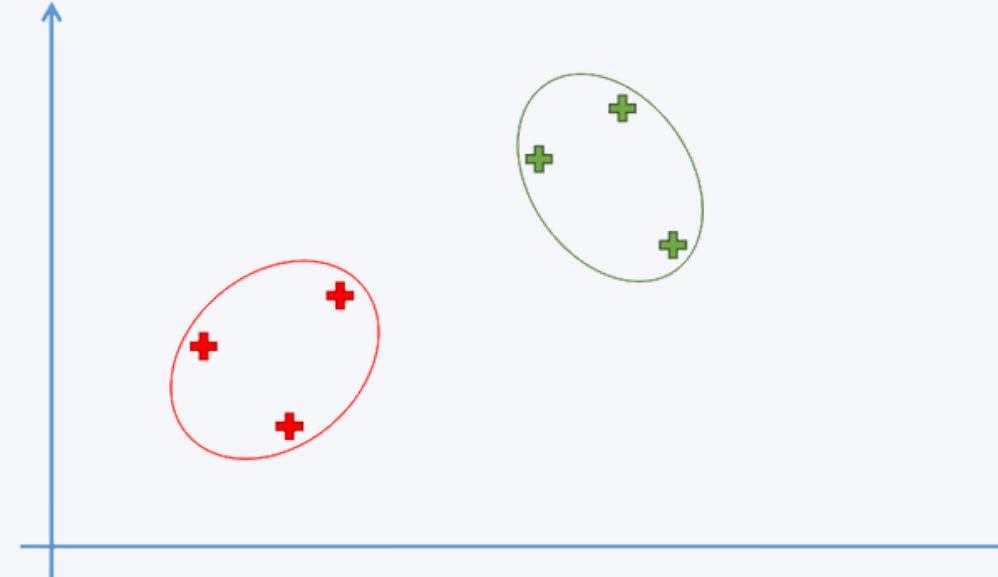
STEP 3: Take the two closest clusters and make them one cluster  
→ That forms 4 clusters



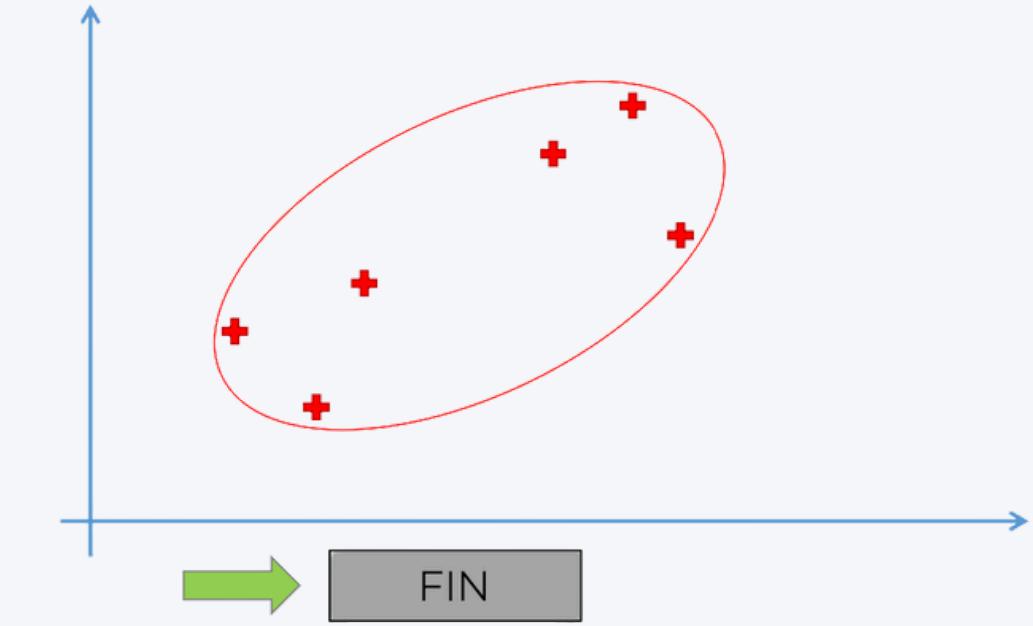
STEP 4: Repeat STEP 3 until there is only one cluster



STEP 4: Repeat STEP 3 until there is only one cluster



STEP 4: Repeat STEP 3 until there is only one cluster

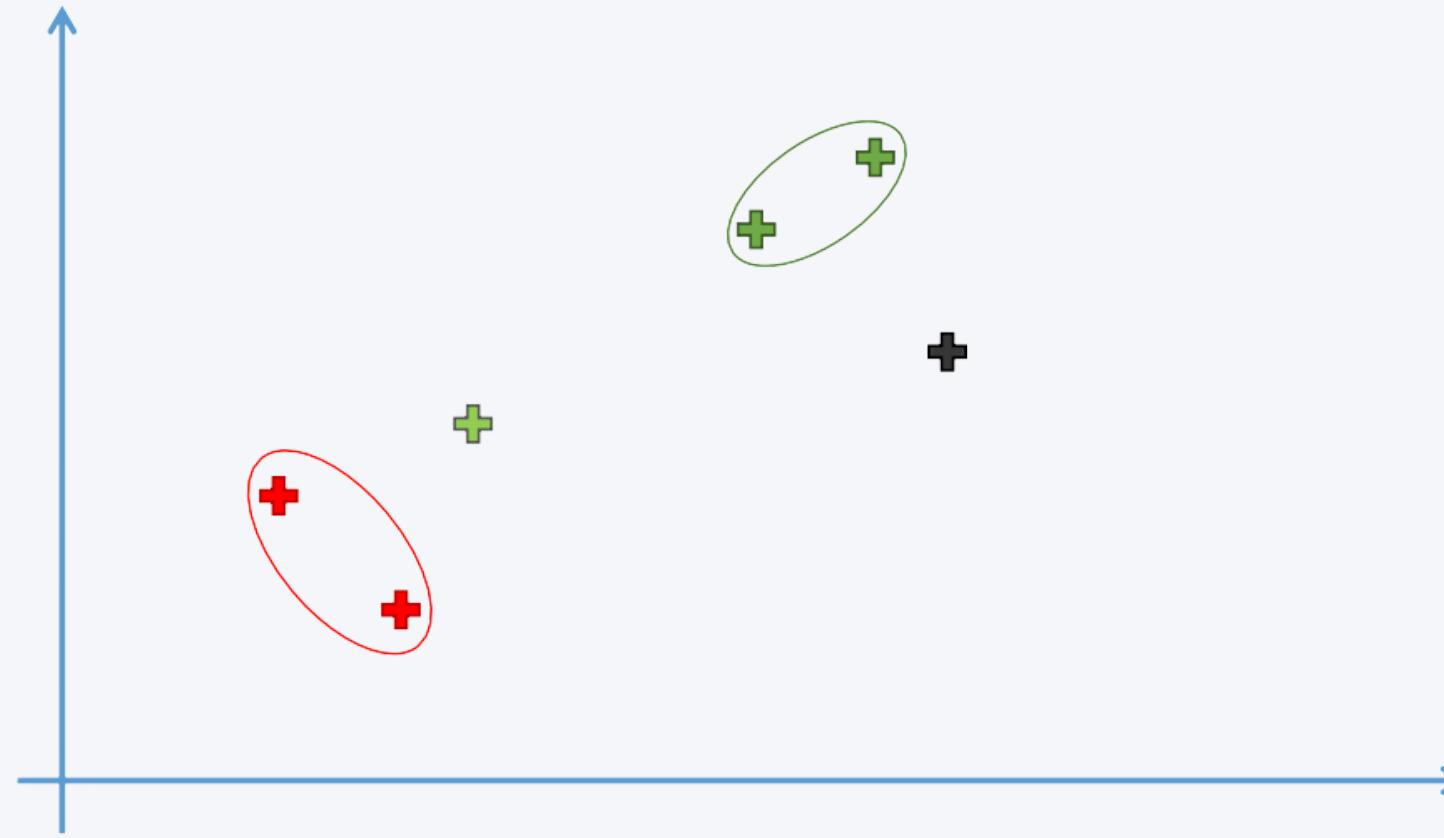


# Hierarchical Clustering

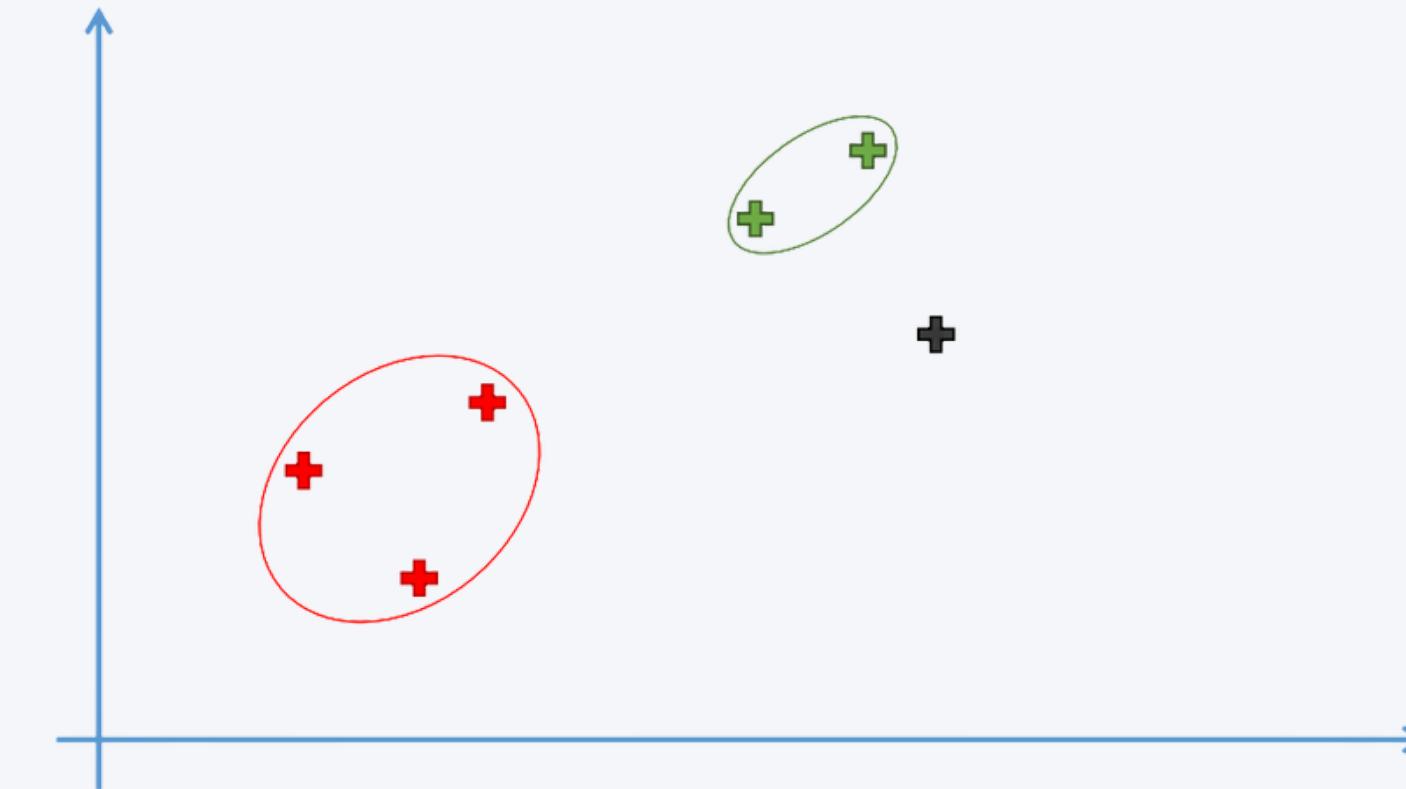
## Agglomerative HC

STEP 3: Take the two closest clusters and make them one cluster

→ That forms 4 clusters



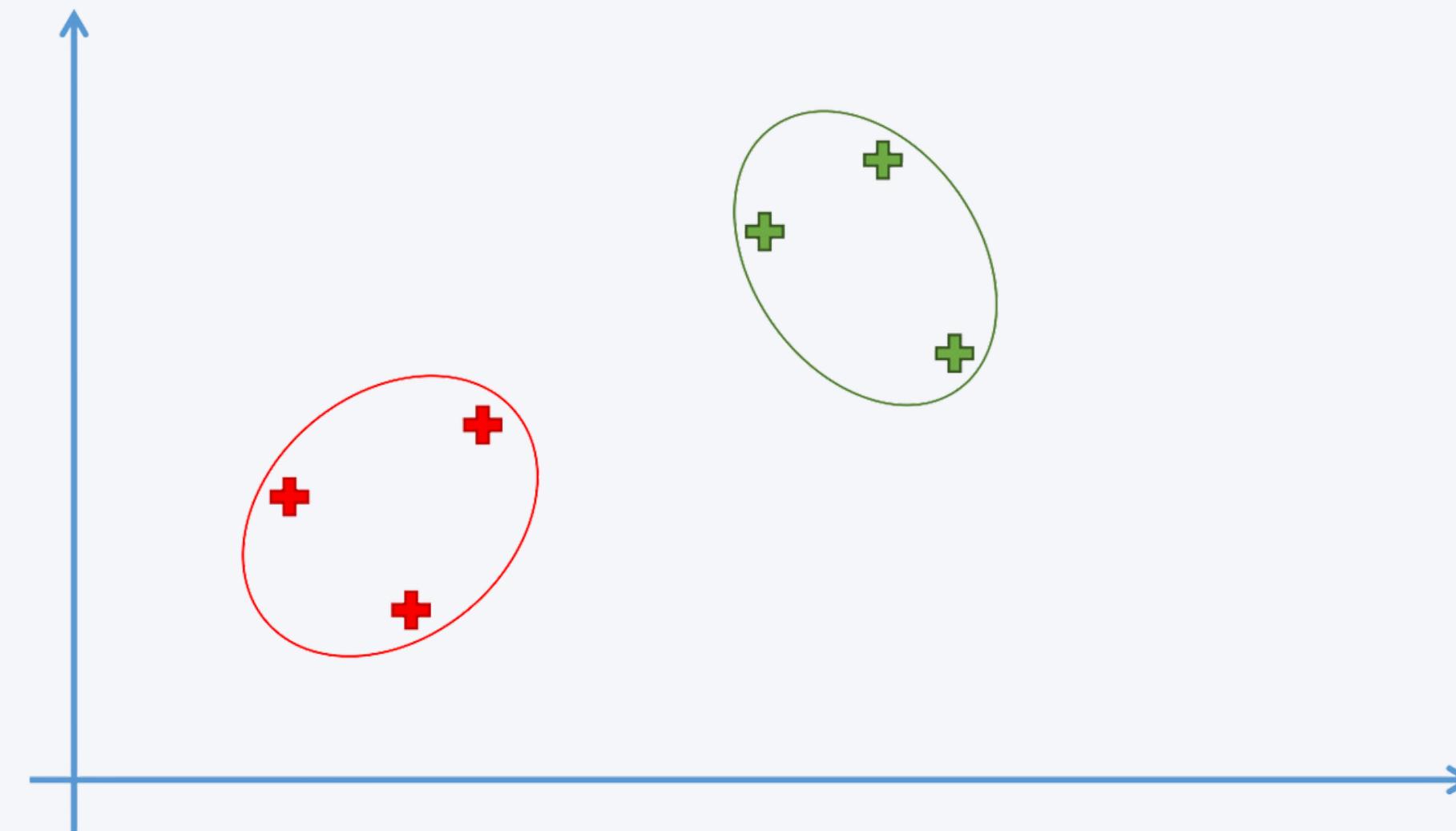
STEP 4: Repeat STEP 3 until there is only one cluster



# Hierarchical Clustering

## Agglomerative HC

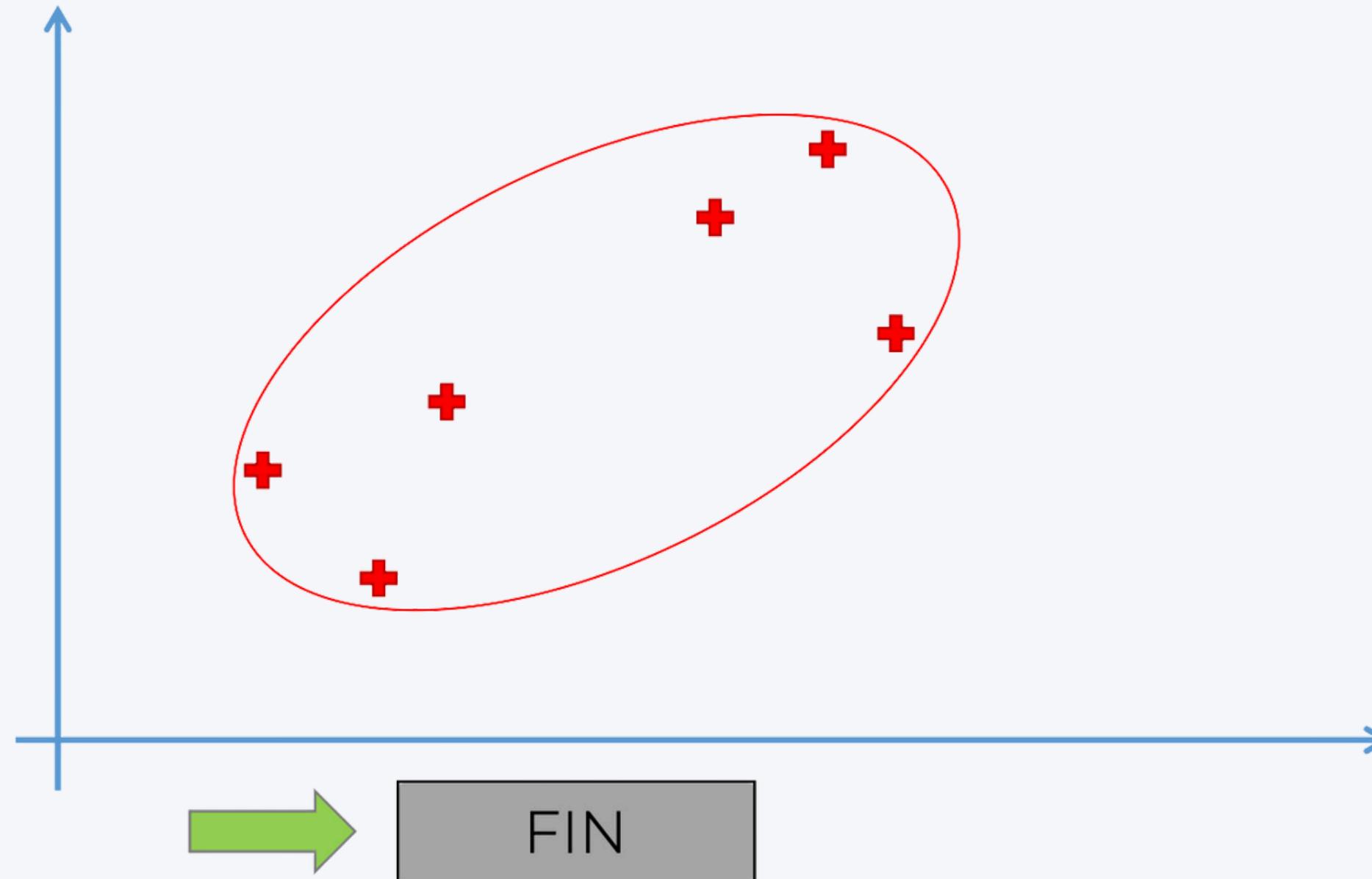
STEP 4: Repeat STEP 3 until there is only one cluster



# Hierarchical Clustering

## Agglomerative HC

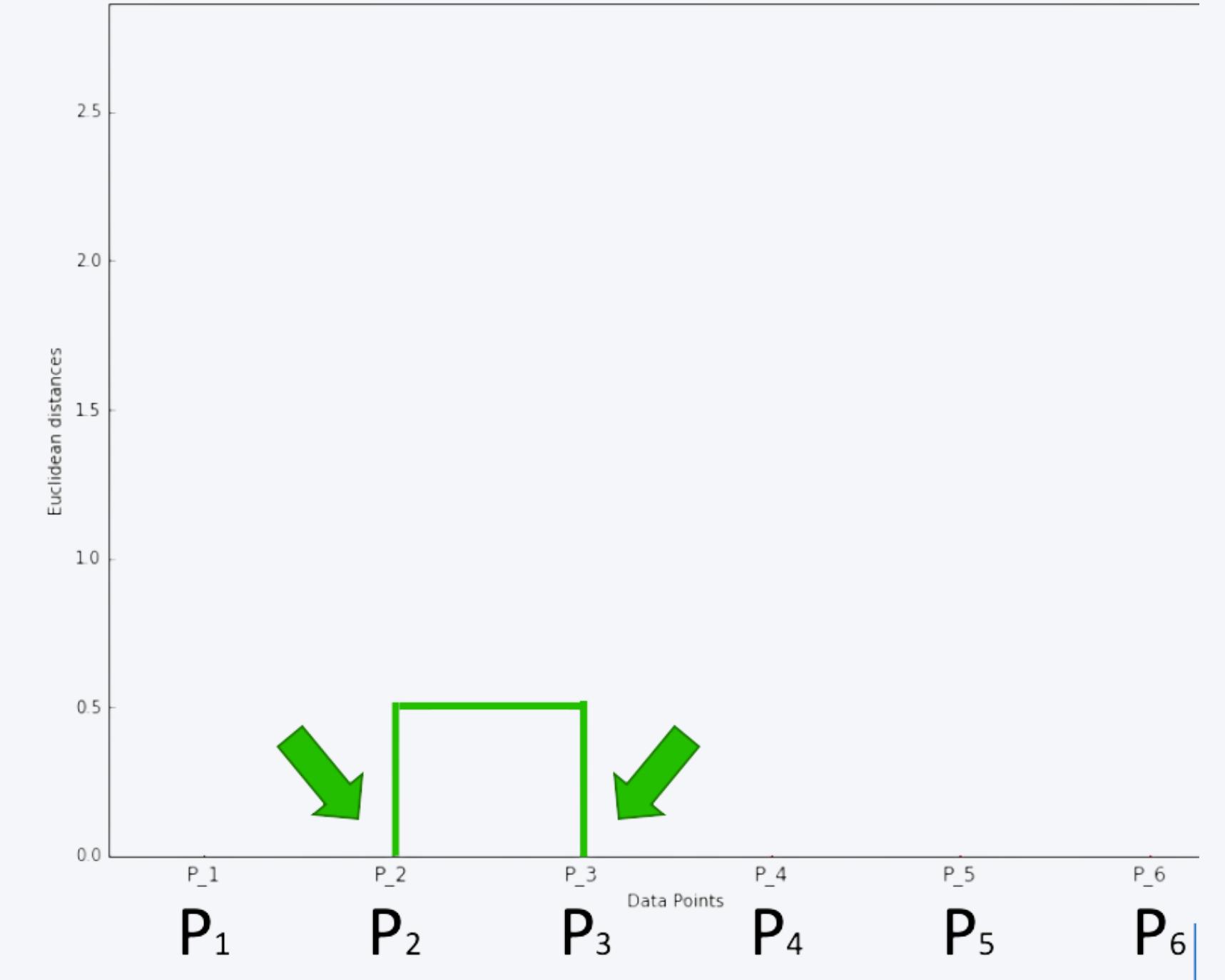
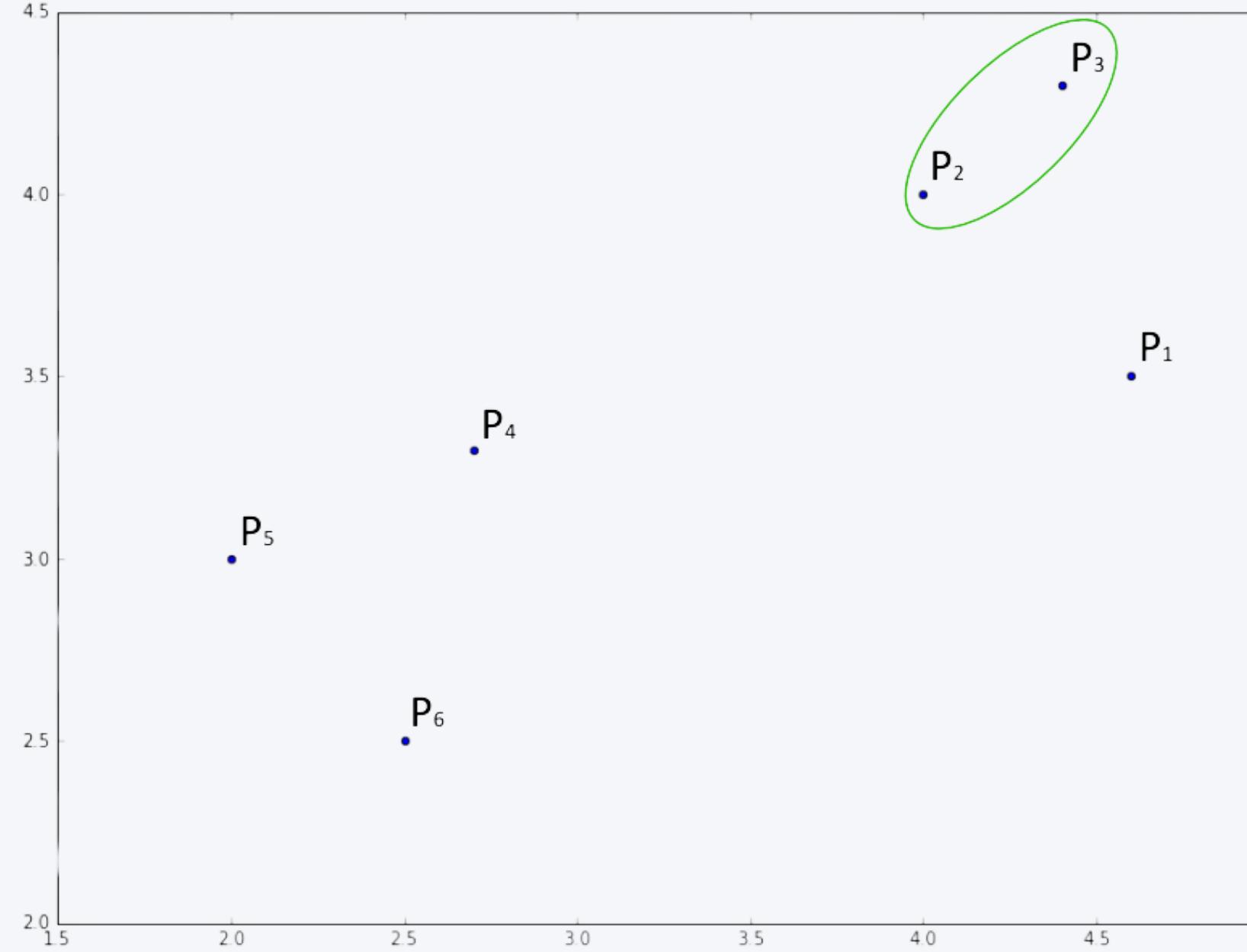
STEP 4: Repeat STEP 3 until there is only one cluster



- it maintains like memory of each steps , saving the steps taken with the distances.

# Hierarchical Clustering

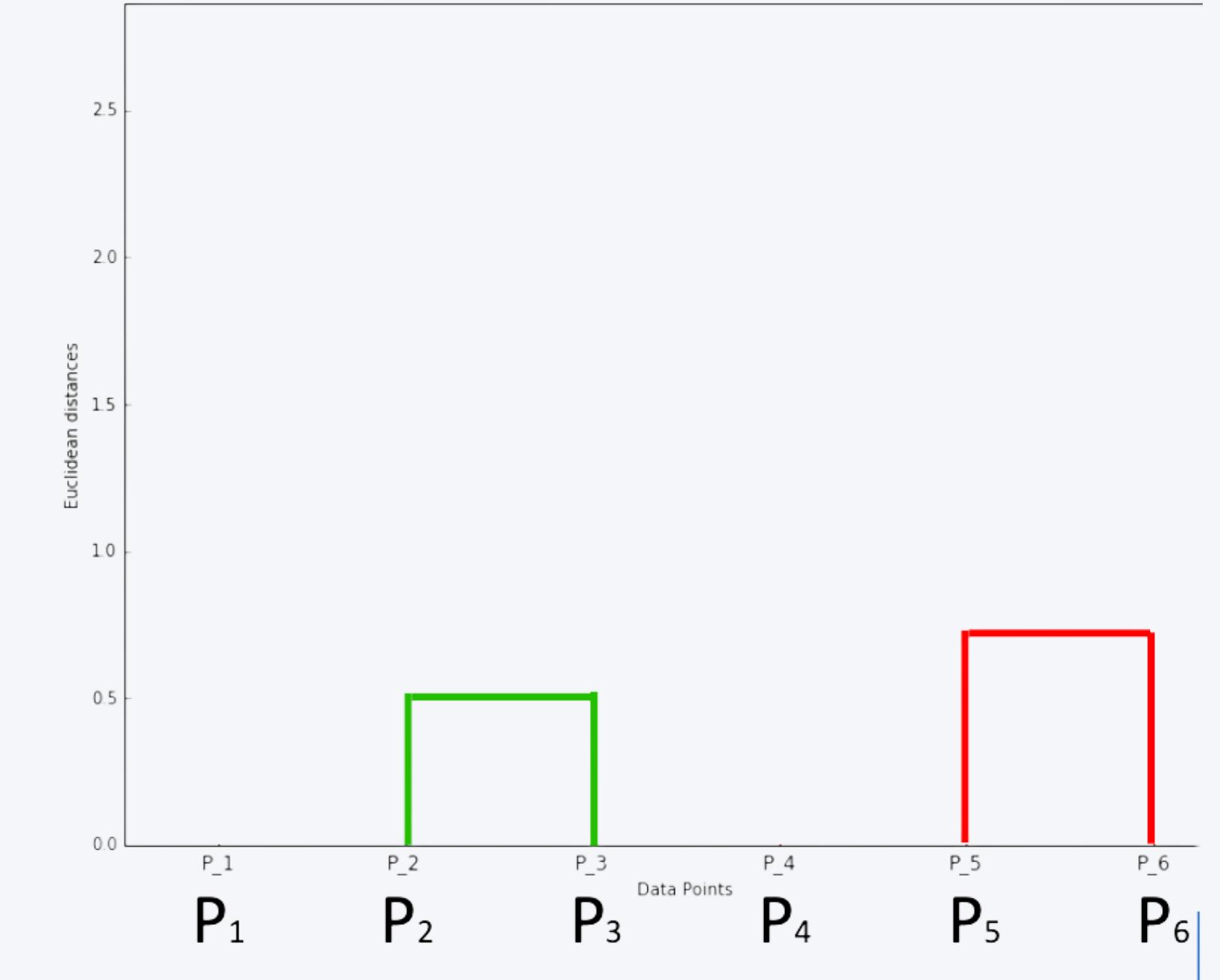
## How Do Dendograms Work ?



- once we defined first closest two points and made clusters, we can now assign p2 and p3 with the distance (Euclidean), and it is the similarity of these clusters

# Hierarchical Clustering

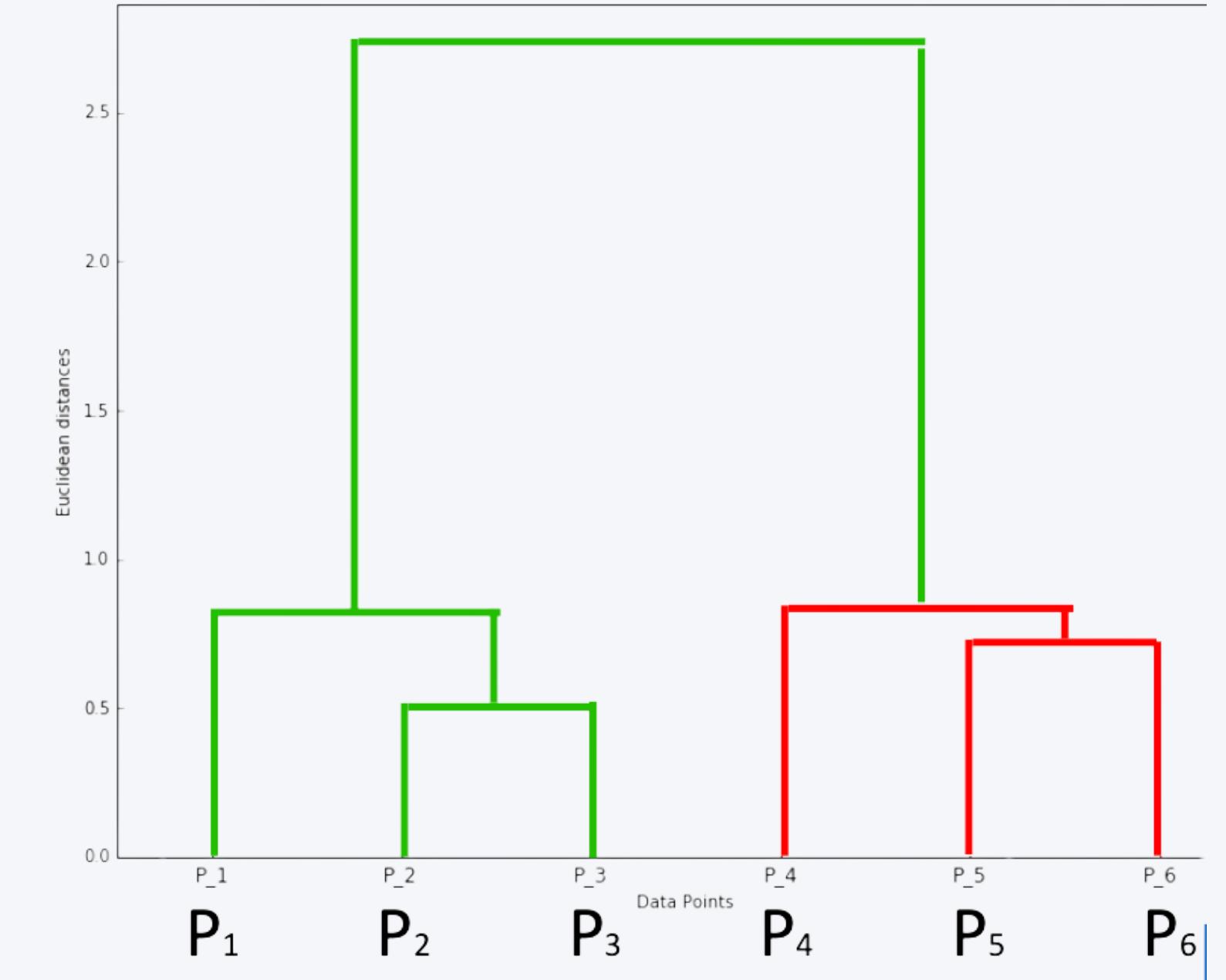
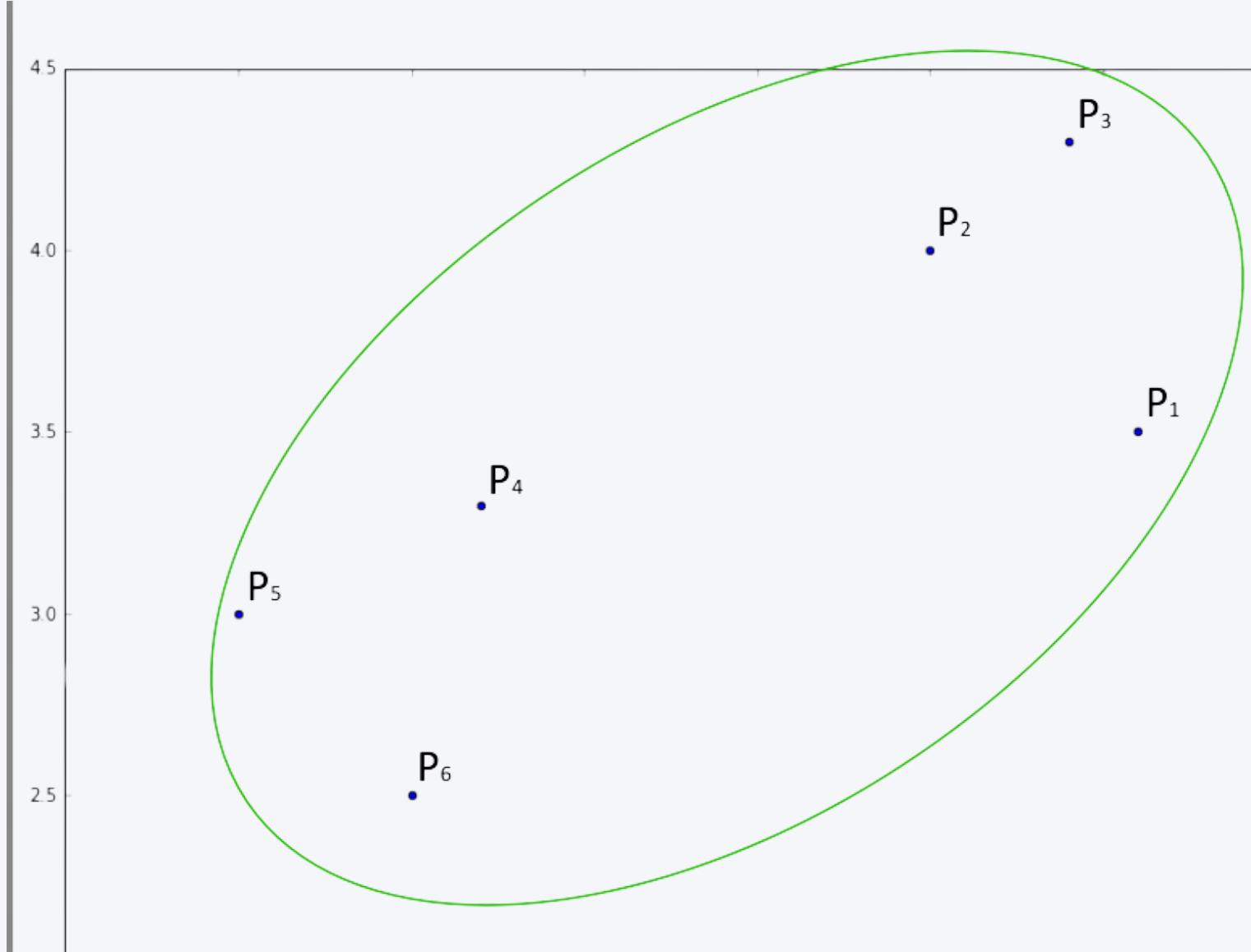
## How Do Dendograms Work ?



- Next step , the next two closest clusters and plot them as Dendograms

# Hierarchical Clustering

## How Do Dendograms Work ?

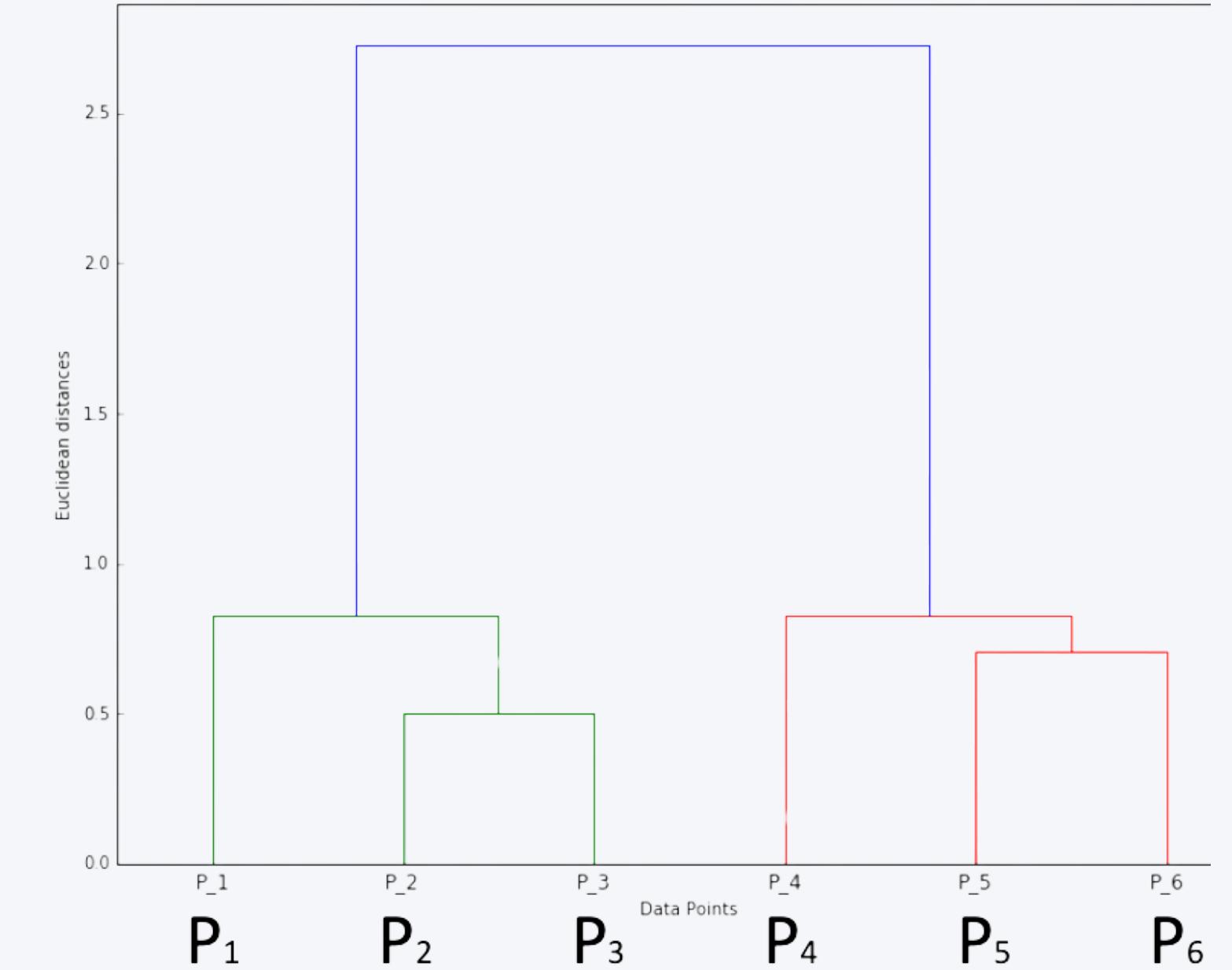
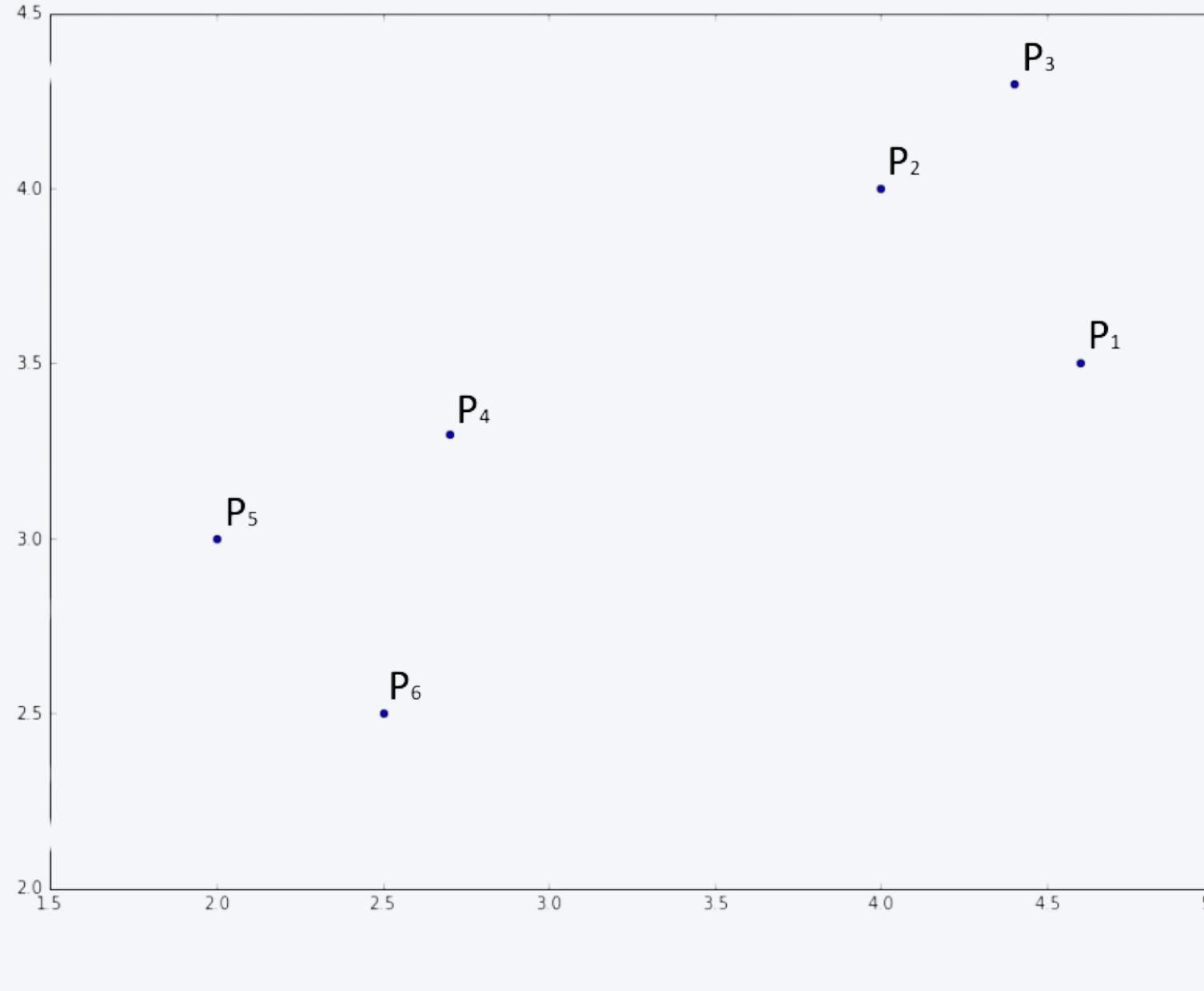


- Repeat the process until you get one cluster

# Hierarchical Clustering



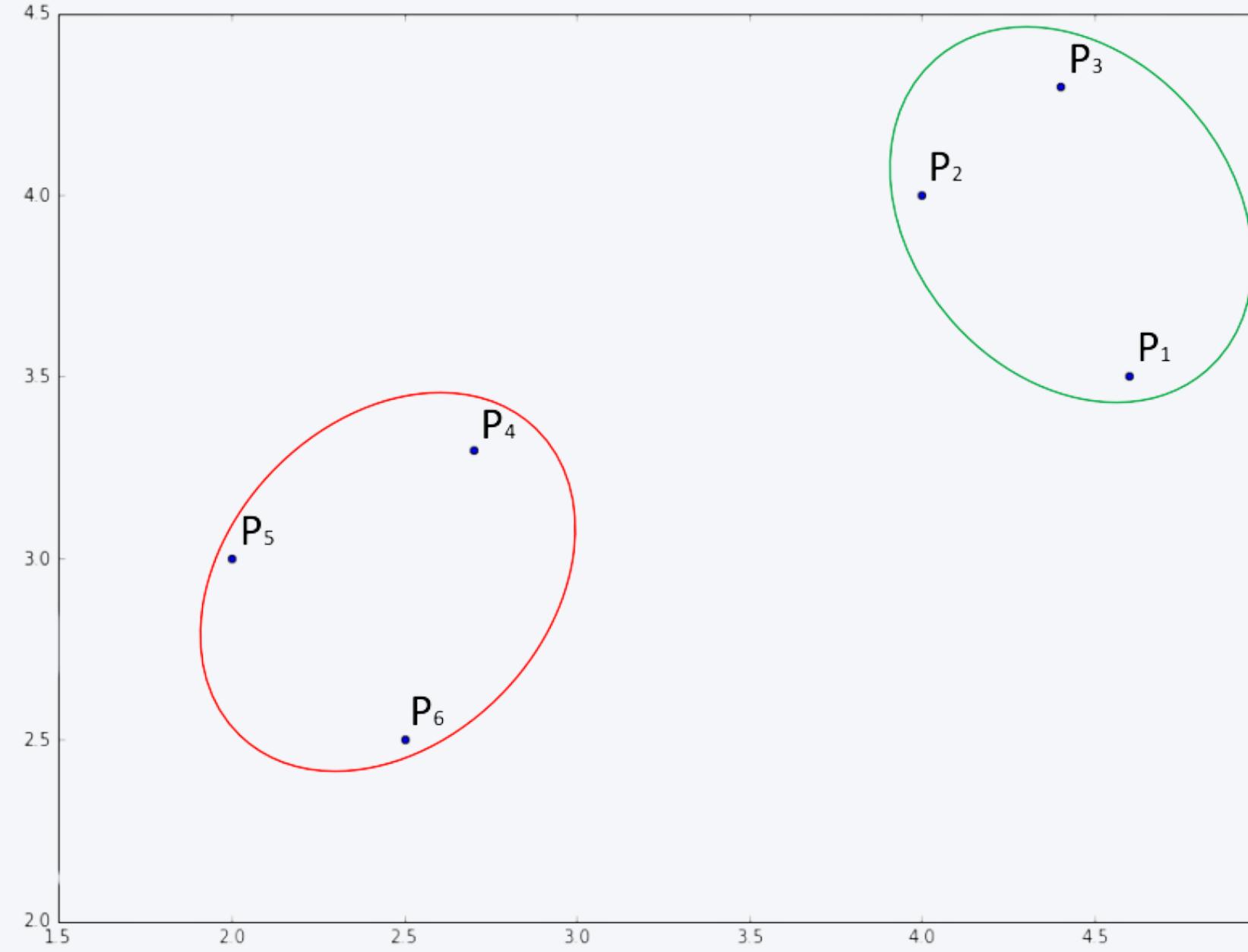
## How Do Dendograms Work ?



- Final Dendograms with clusters and distances (Memory on how clusters are formed)

# Hierarchical Clustering

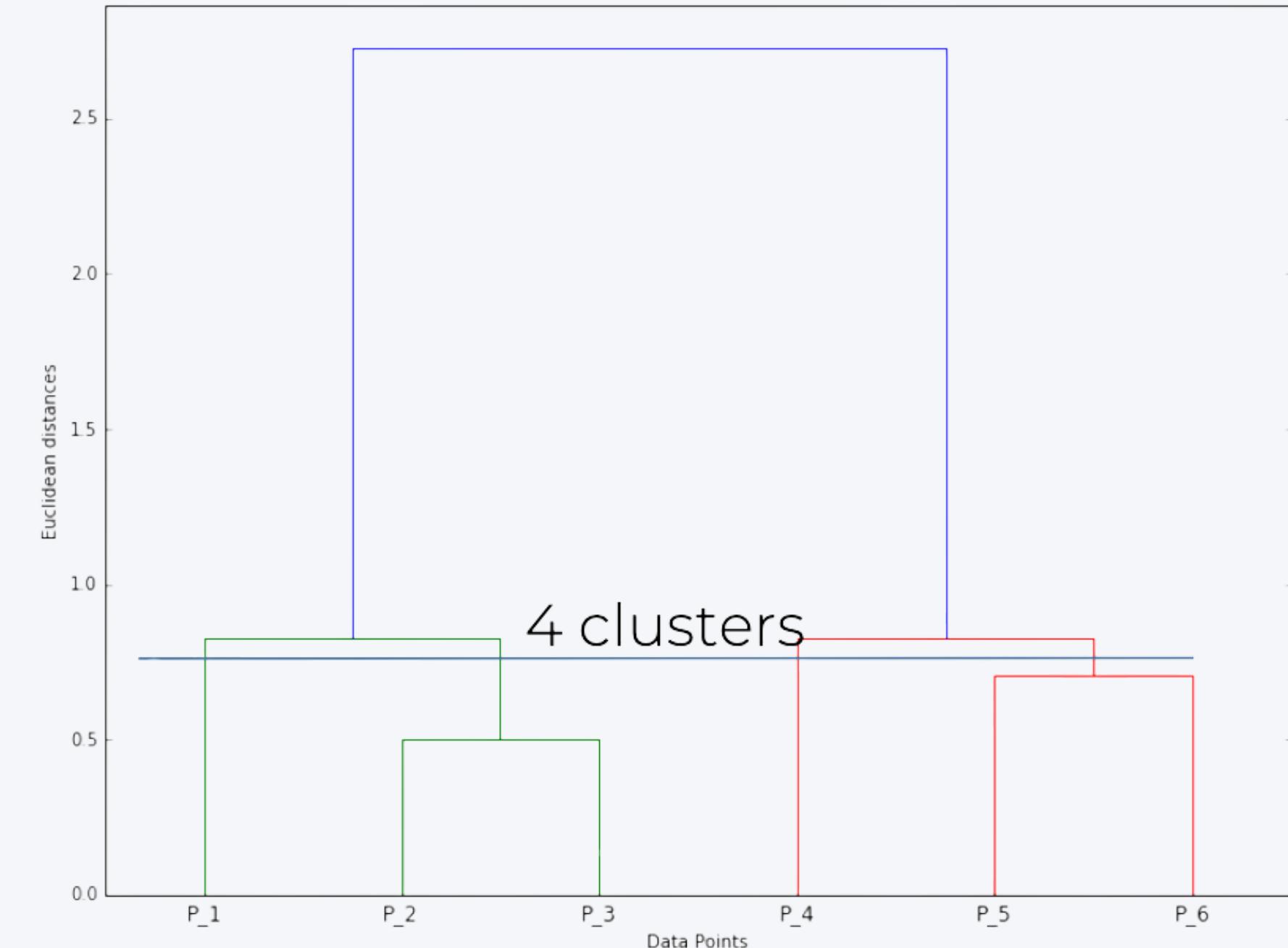
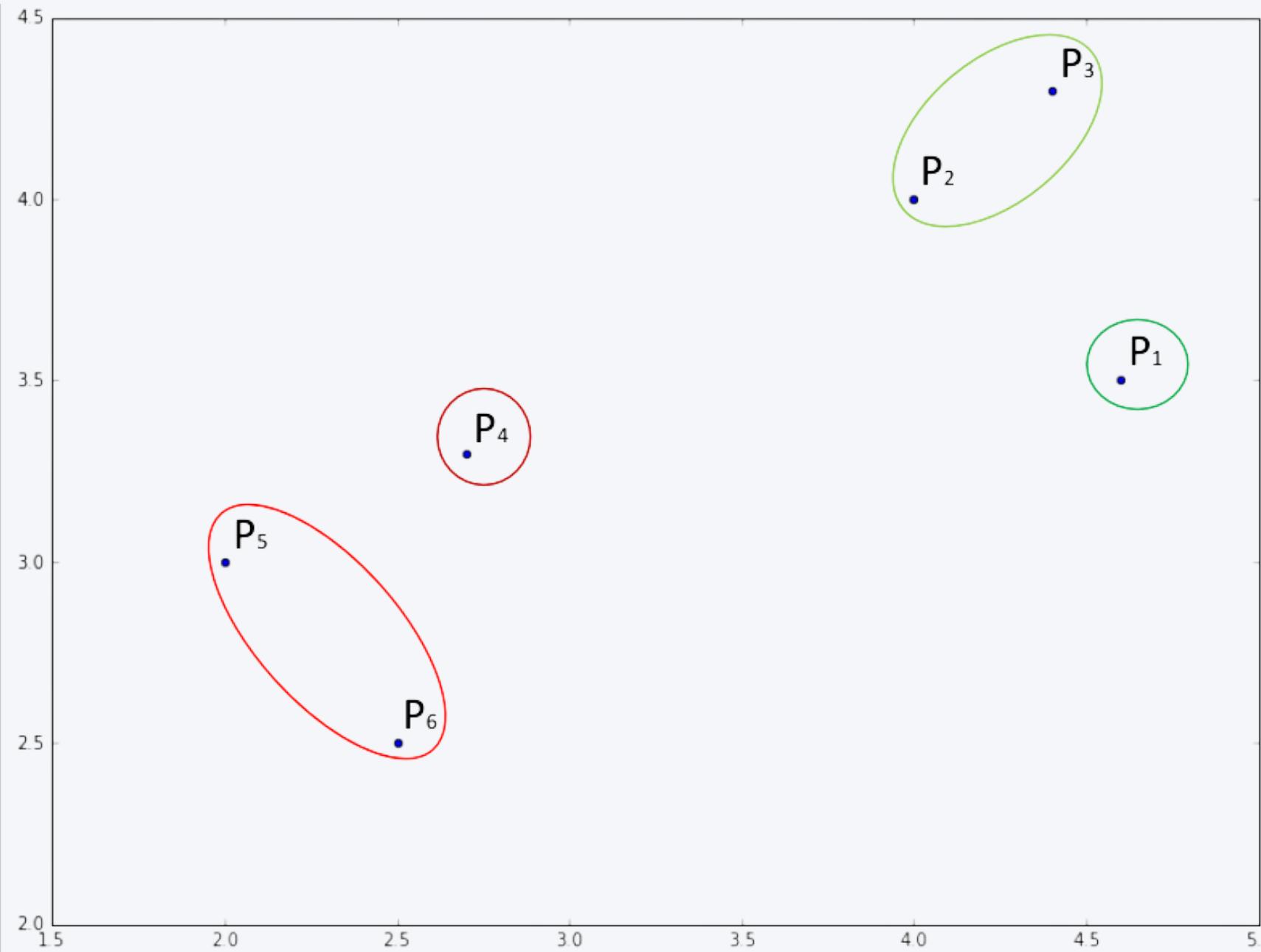
## Using Dendograms ?



- We can set threshold (distance threshold or dissimilarities) which help in making better clusters // by vertical lines, we can see how many clusters we have.

# Hierarchical Clustering

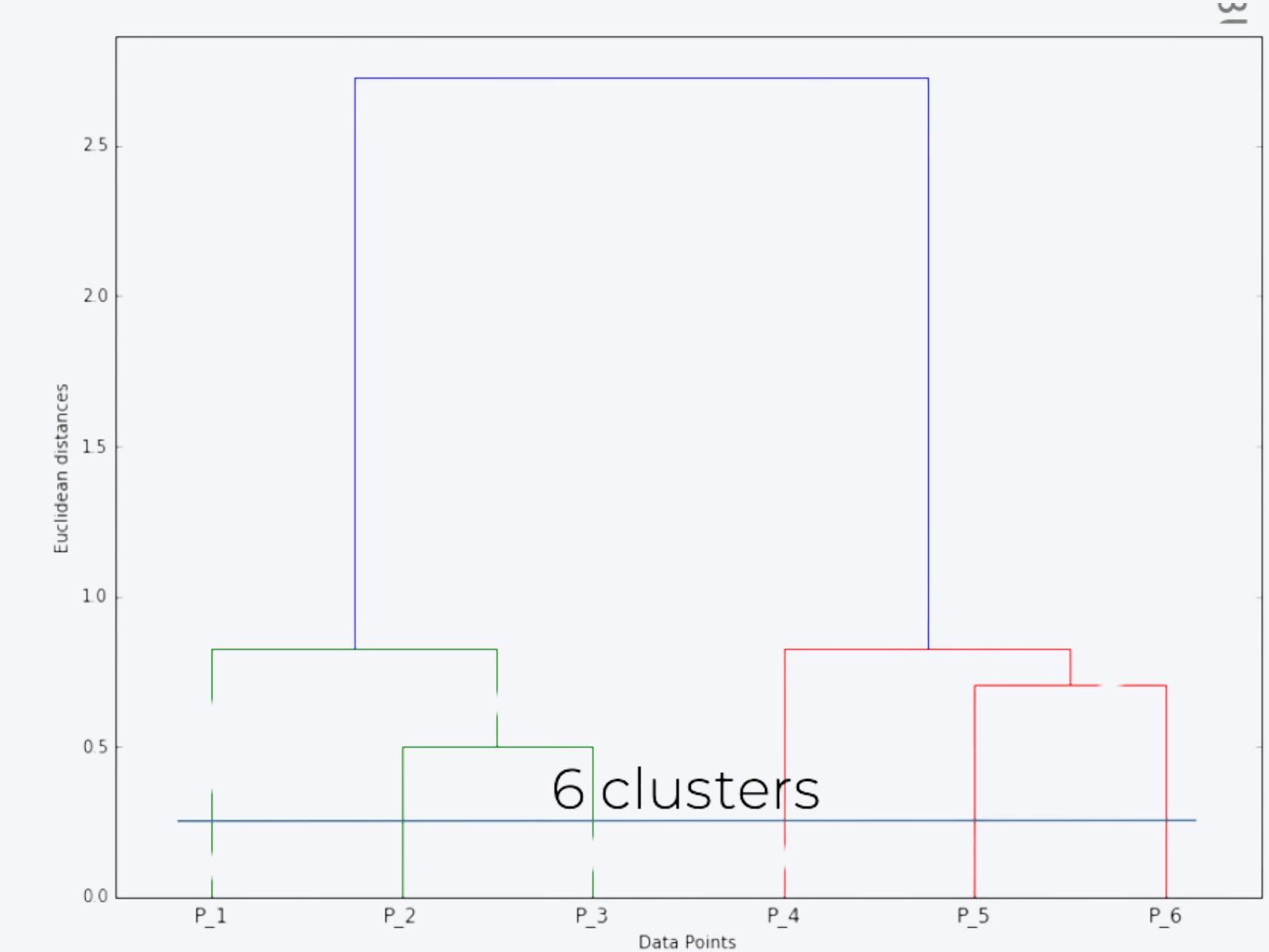
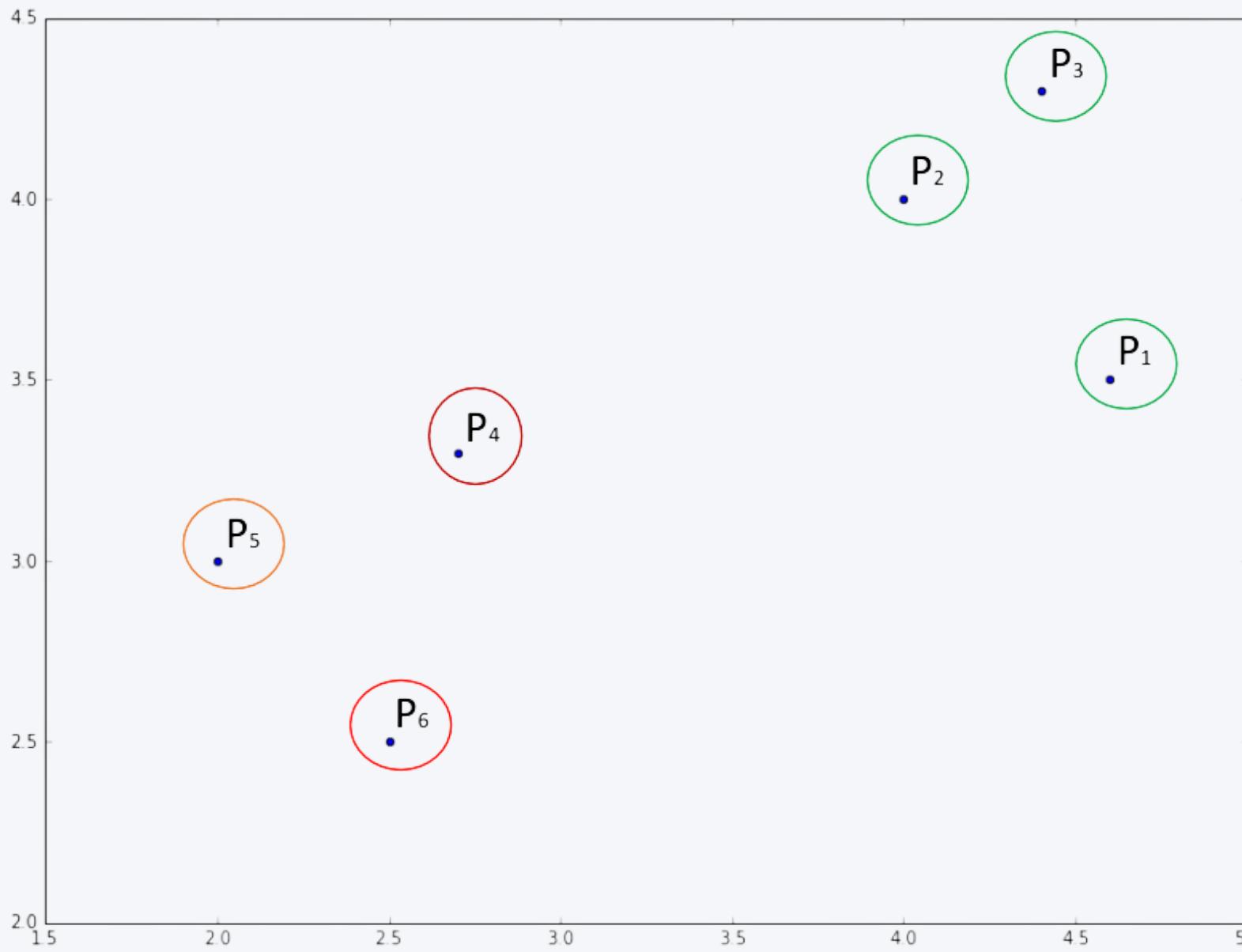
## Using Dendograms ?



- Another threshold

# Hierarchical Clustering

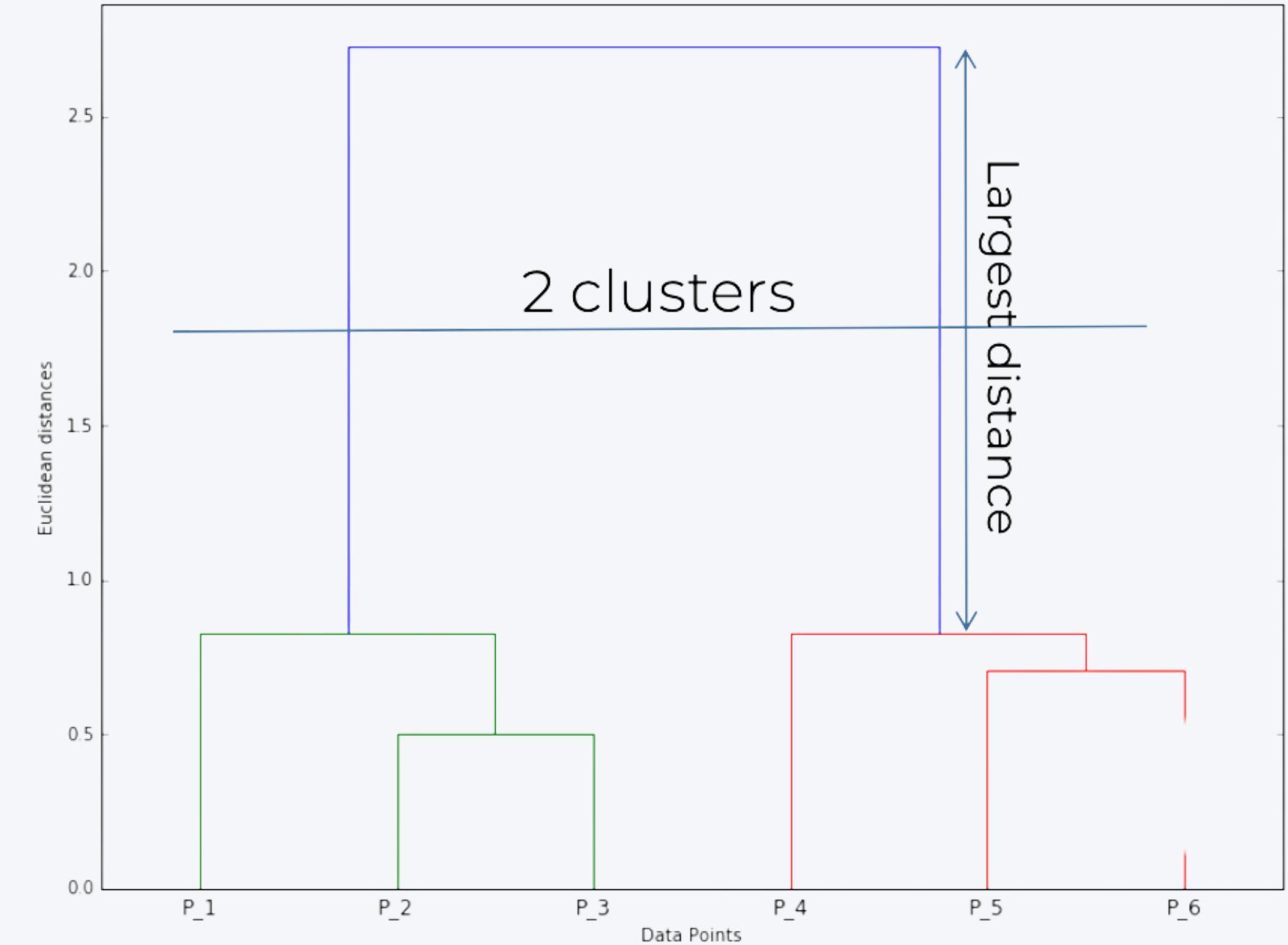
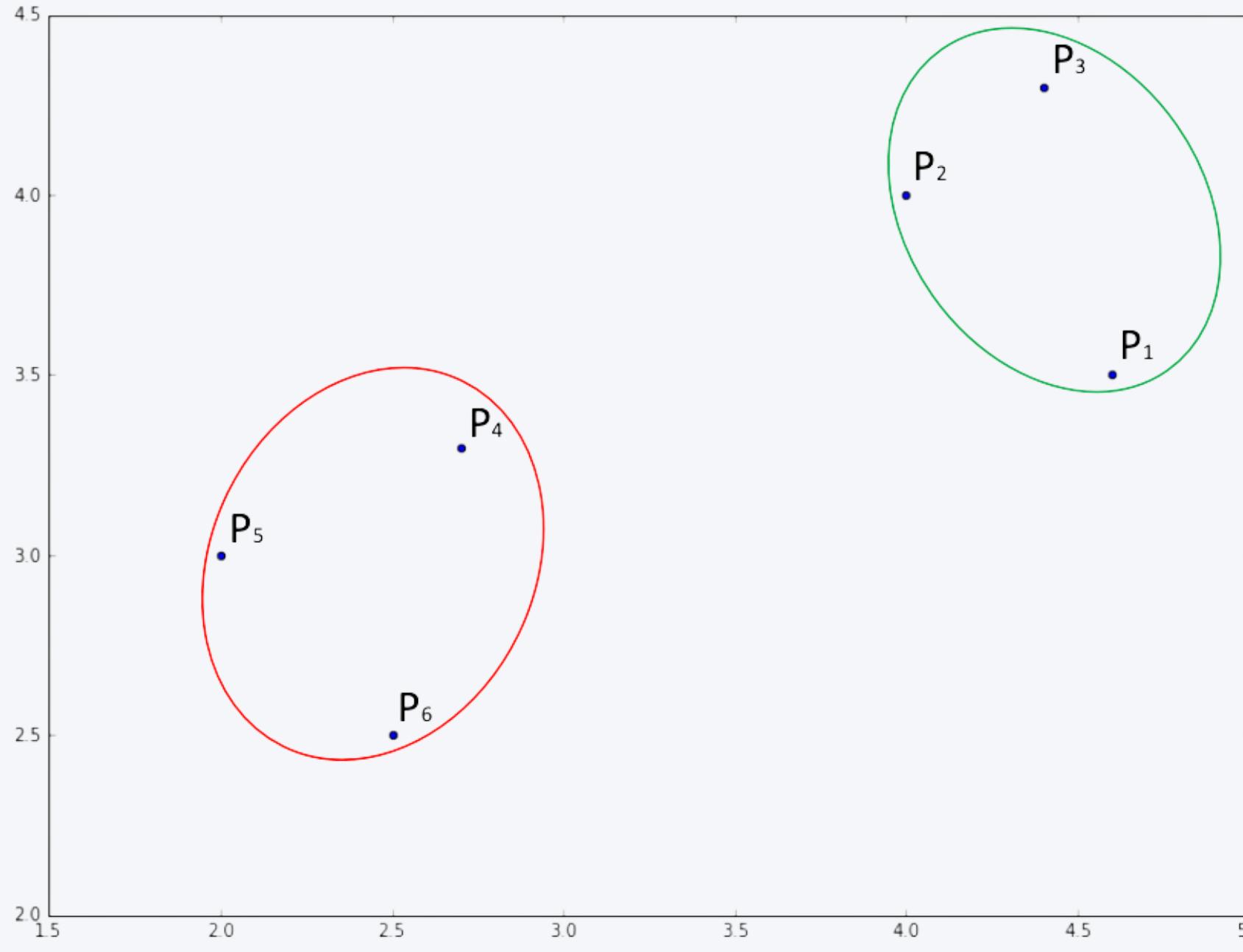
## Using Dendograms ?



- Threshold below our first clusters

# Hierarchical Clustering

## Using Dendograms | Optimal Number of clusters:

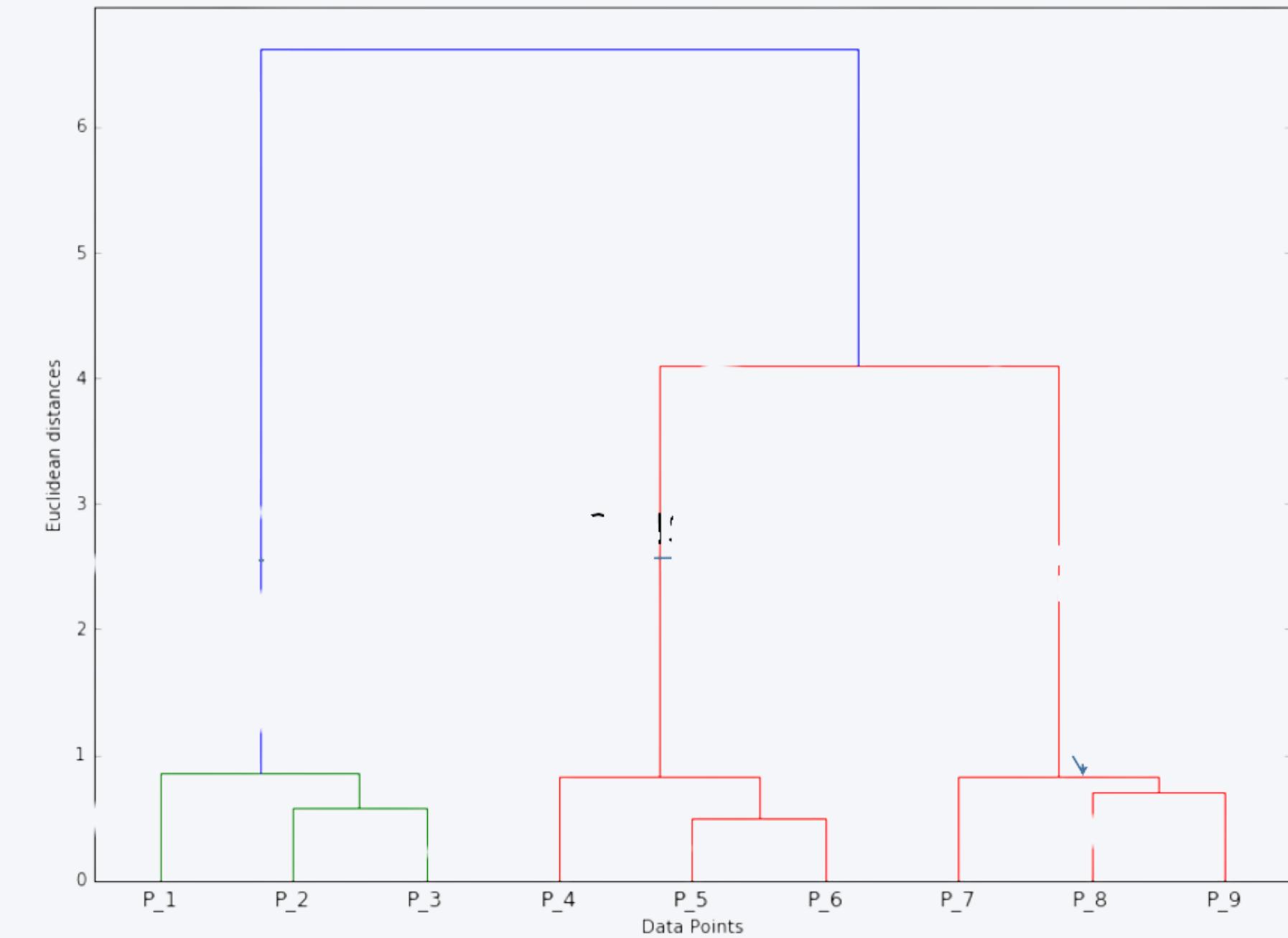
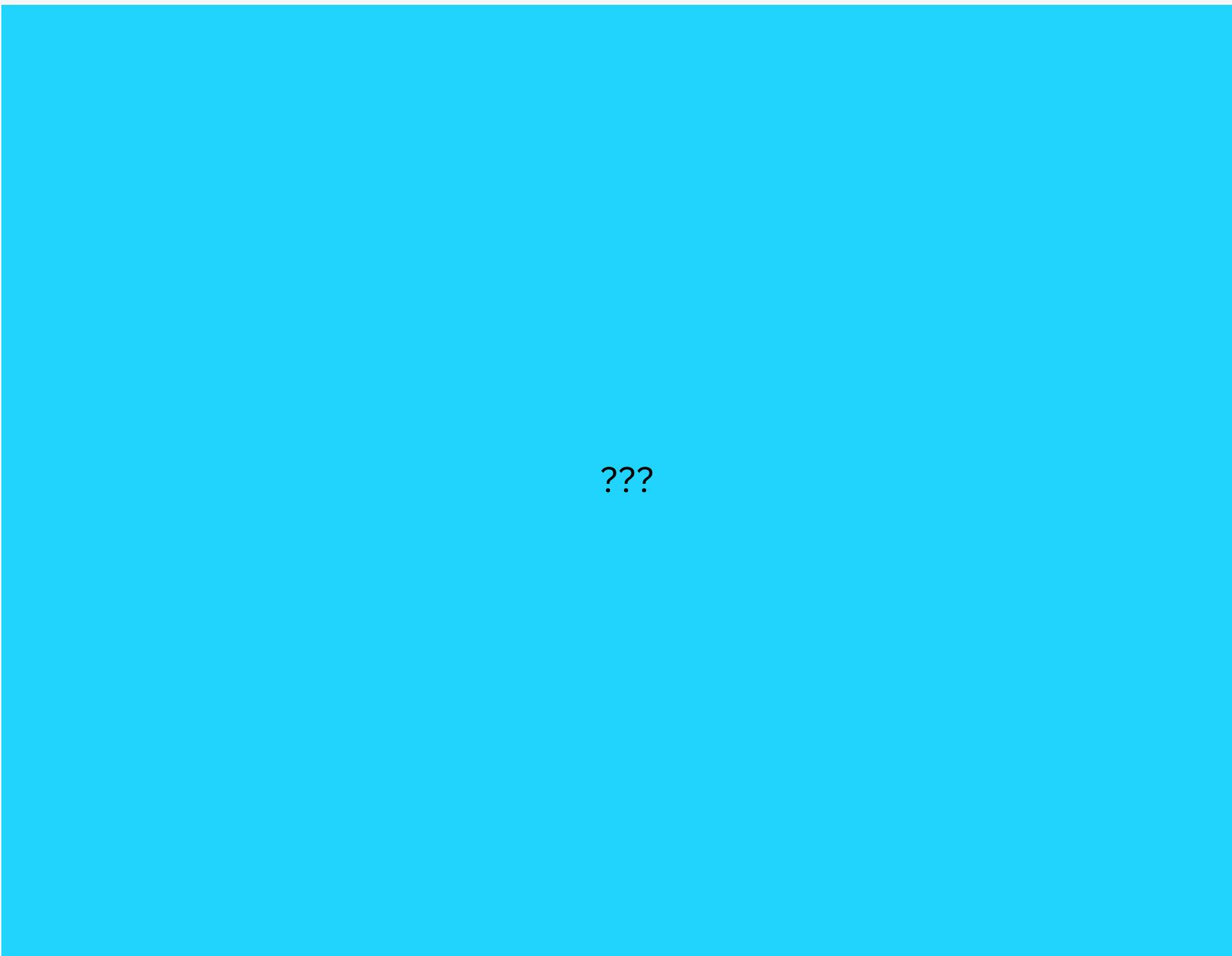


- By using vertical distance that **not** cross any horizontal line and should be **longest** line
-

# Hierarchical Clustering

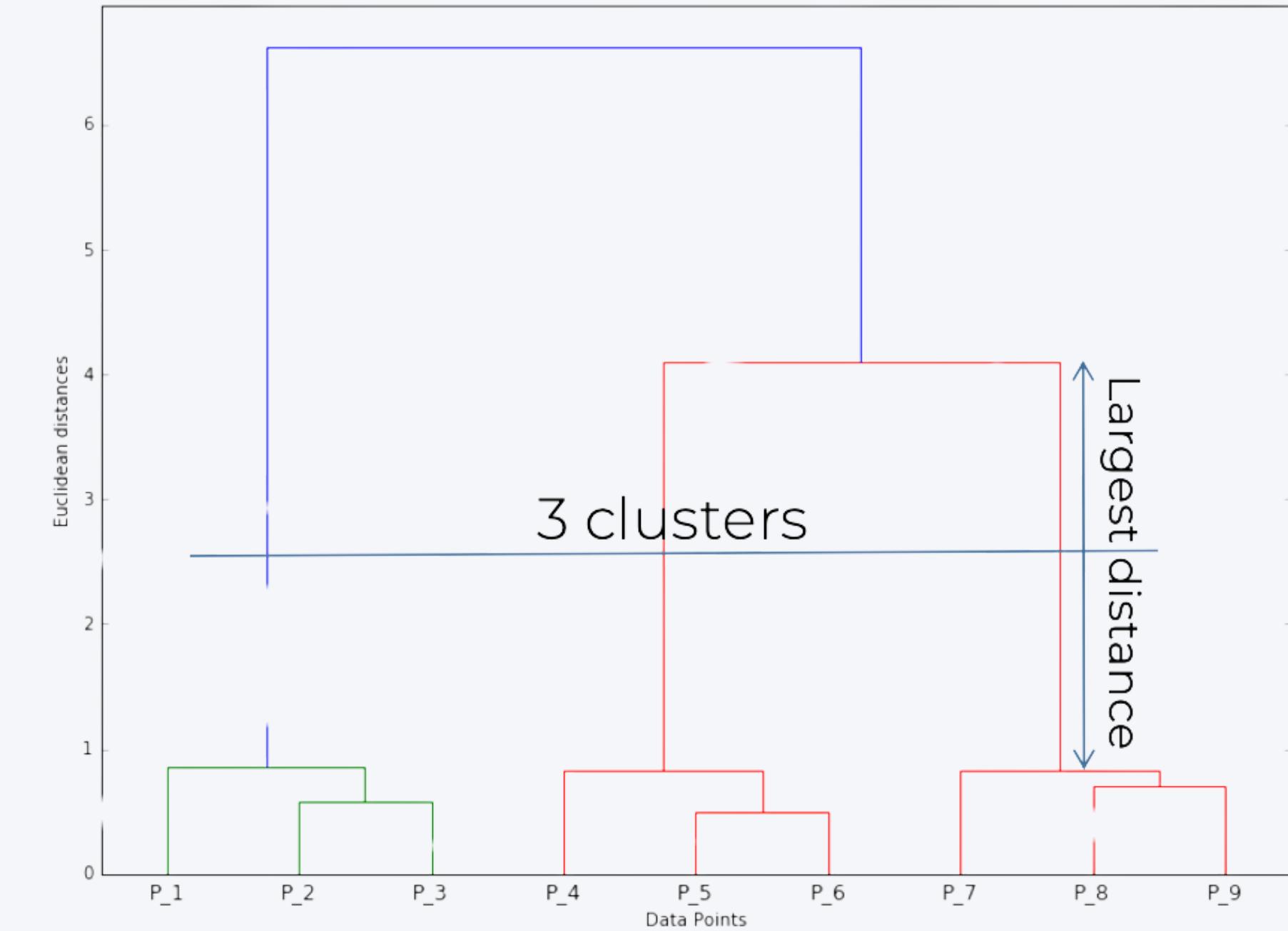
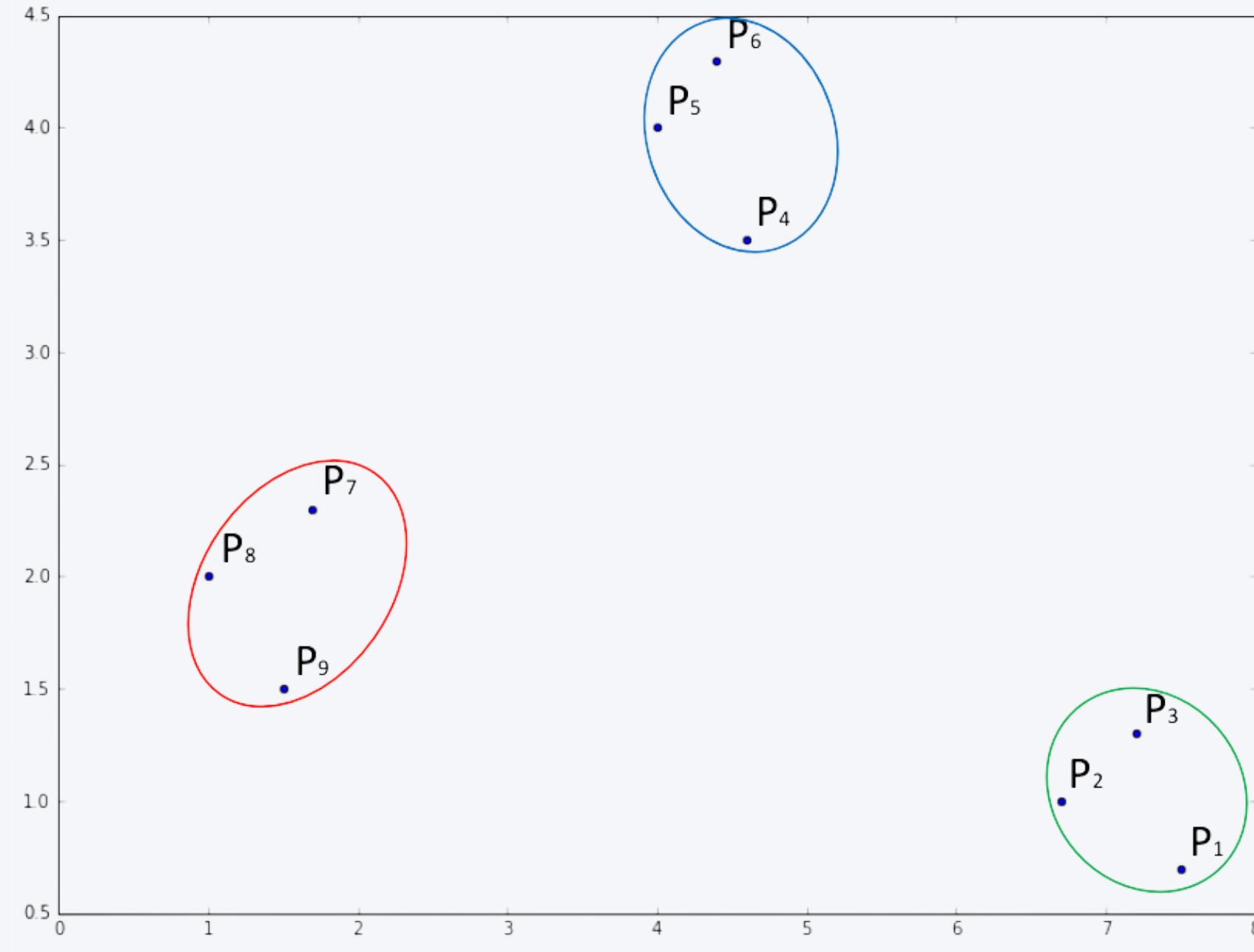


## Dendograms | Knowledge Test



# Hierarchical Clustering

## Dendograms | Knowledge Test



# Clustering

## Difference Between Clustering Models

Clustering Model	Pros	Cons
K-Means	Simple to understand, easily adaptable, works well on small or large datasets, fast, efficient and performant	Need to choose the number of clusters
Hierarchical Clustering	The optimal number of clusters can be obtained by the model itself, practical visualisation with the dendrogram	Not appropriate for large datasets

# Hands-On Code

Clustering Implementation