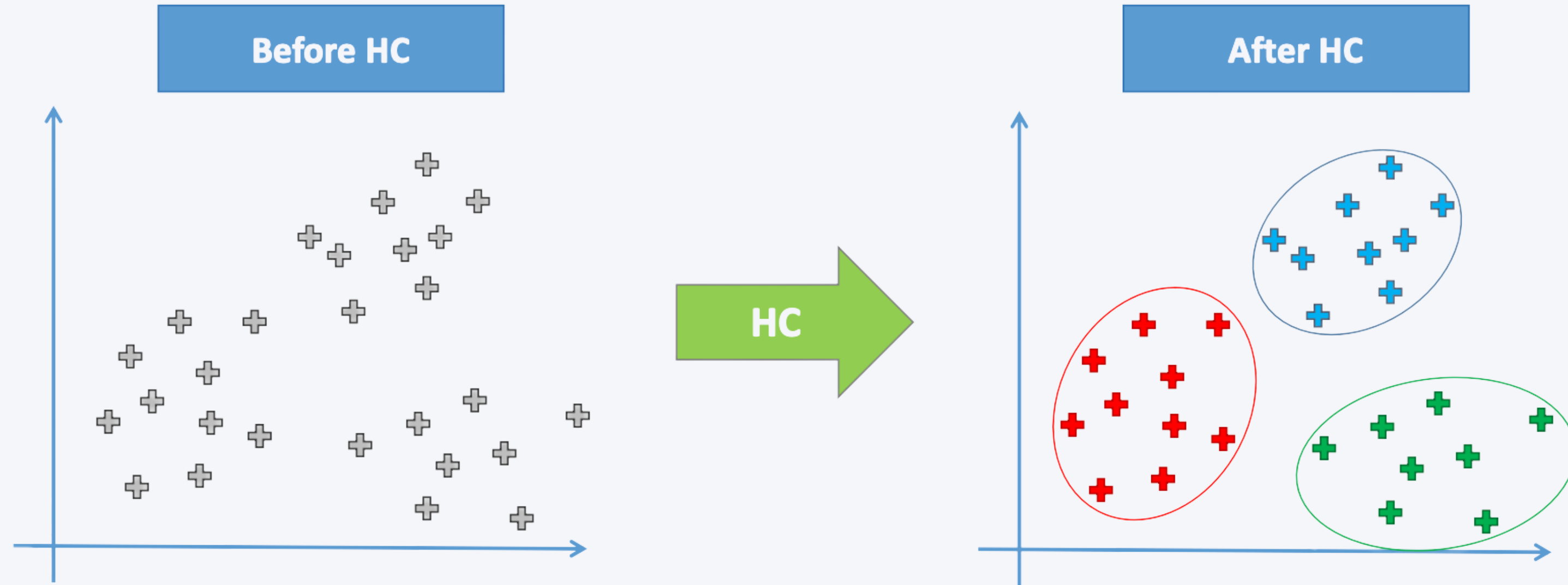


Hierarchical CLustering

Hierarchical CLustering

What HC does ?



Same as K-Means but different process

Two types of HC:

Agglomerative & Divisive

Bottom-Up Approach



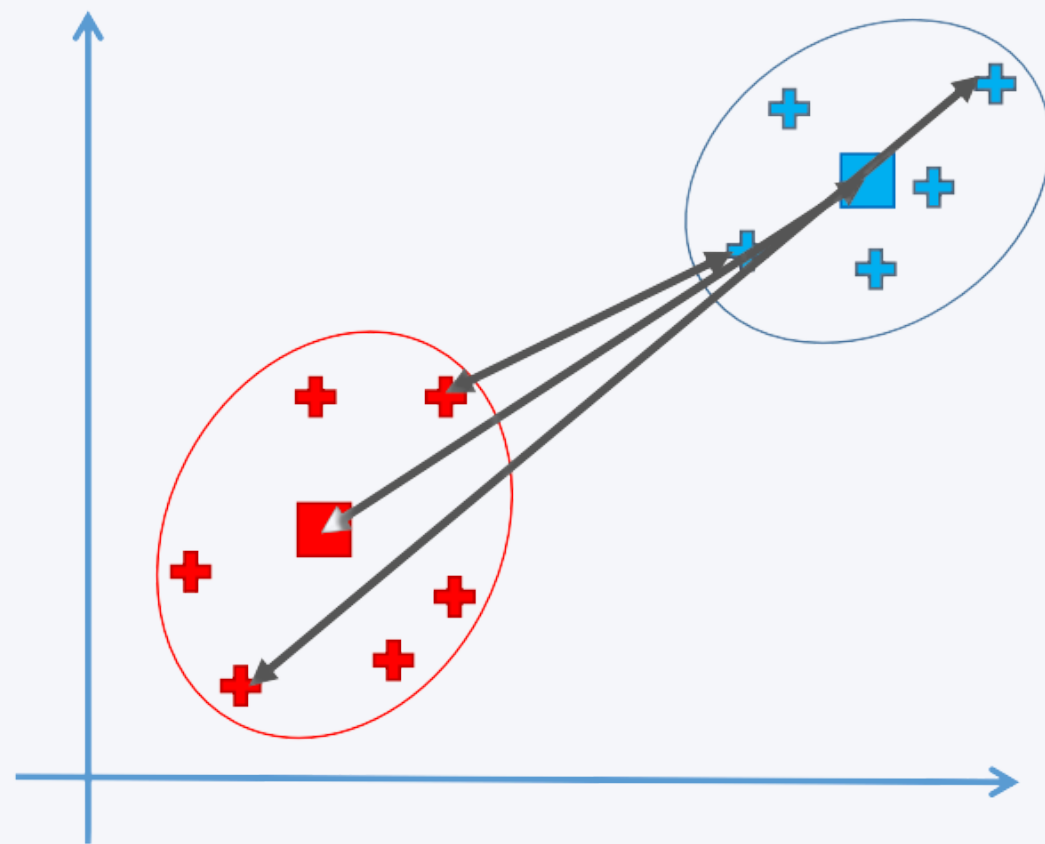
Agglomerative HC

- **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**



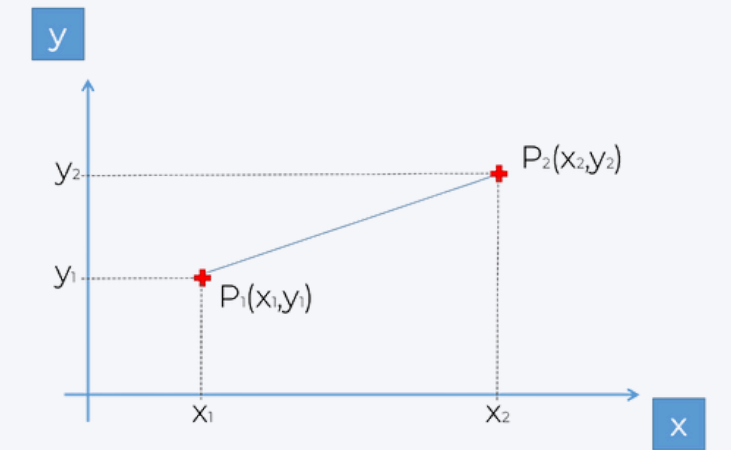
Hierarchical Clustering

Agglomerative HC



Distance Between Two Clusters:

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



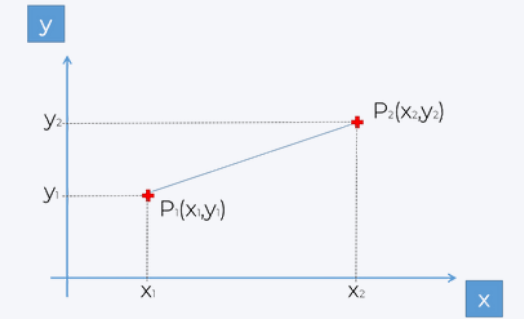
$$\text{Euclidean Distance between } P_1 \text{ and } P_2 = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

- 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** !!

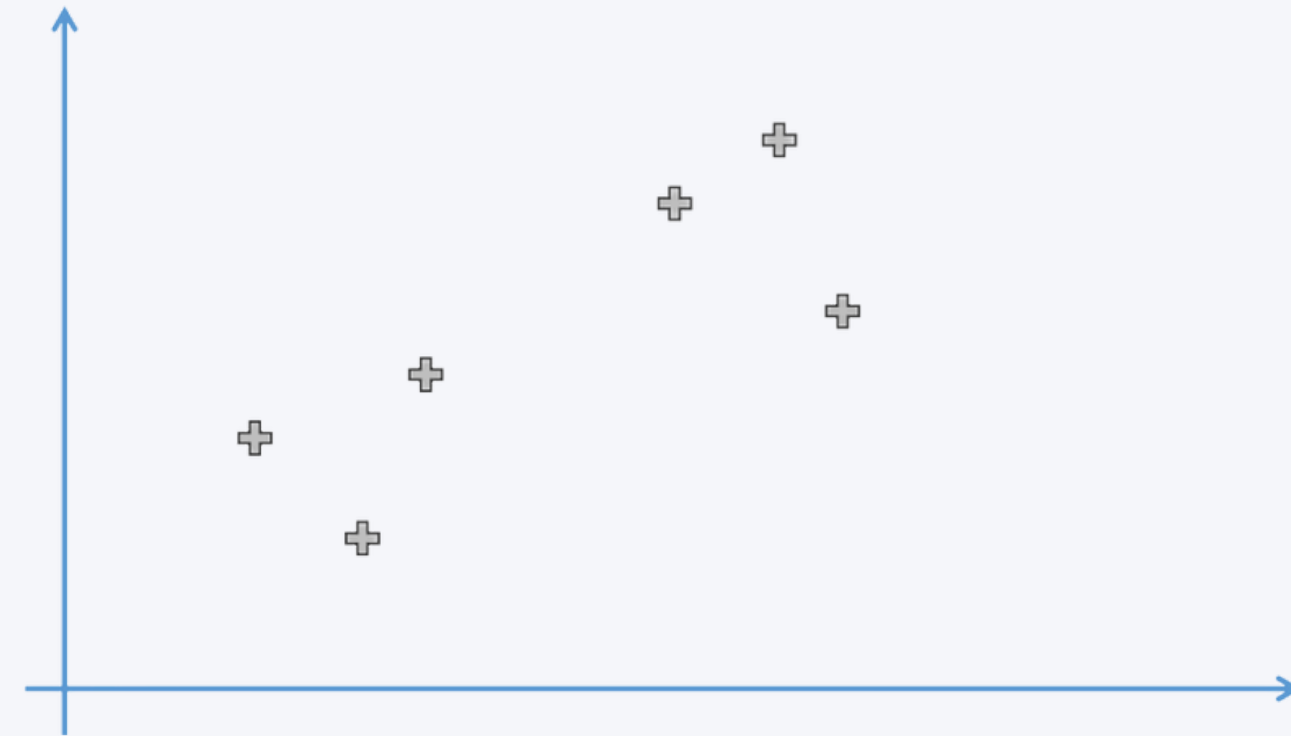
Hierarchical Clustering

Agglomerative HC



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

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

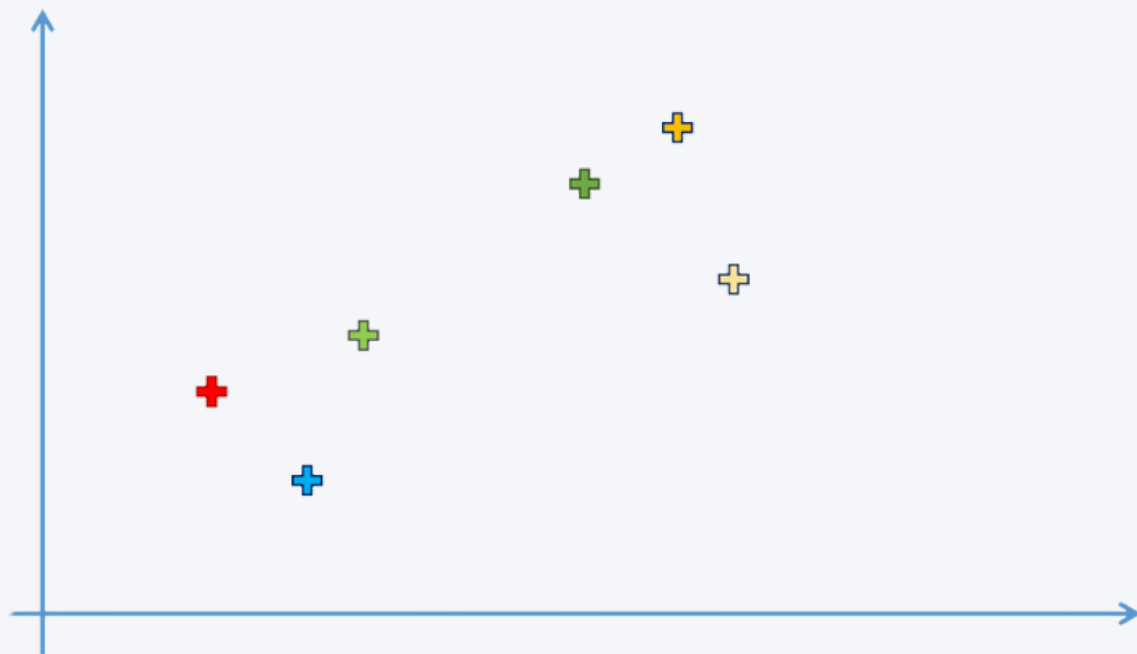


Hierarchical Clustering

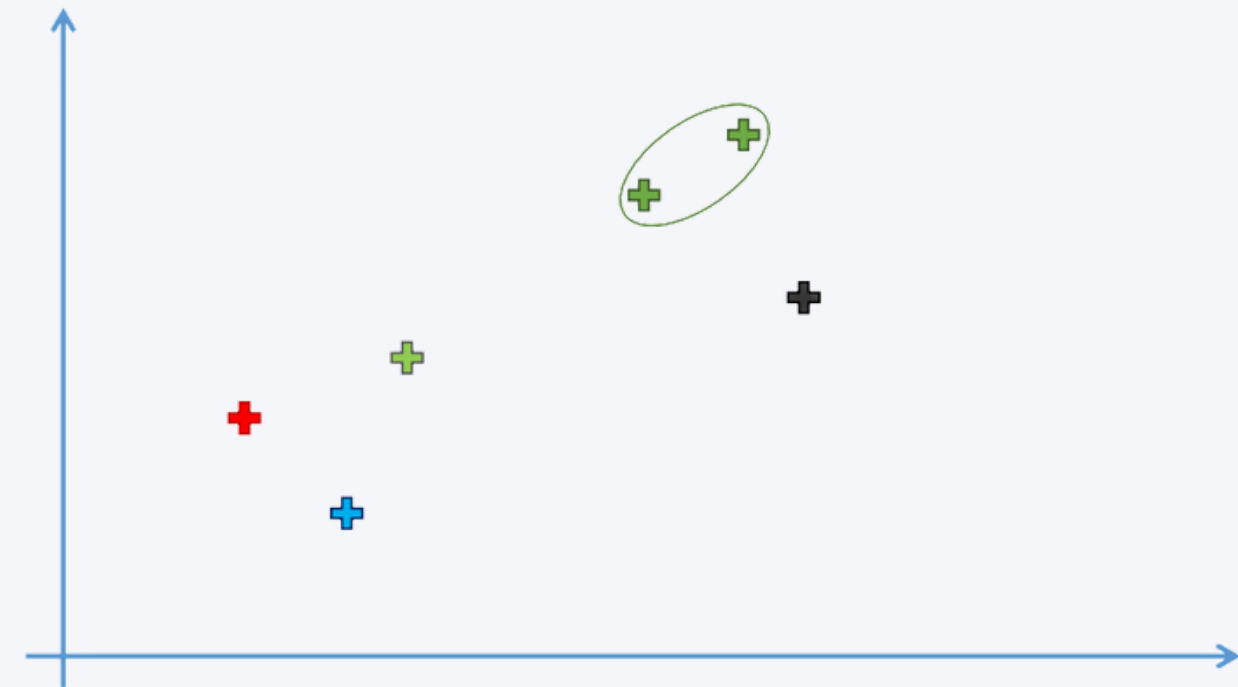


Agglomerative HC

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



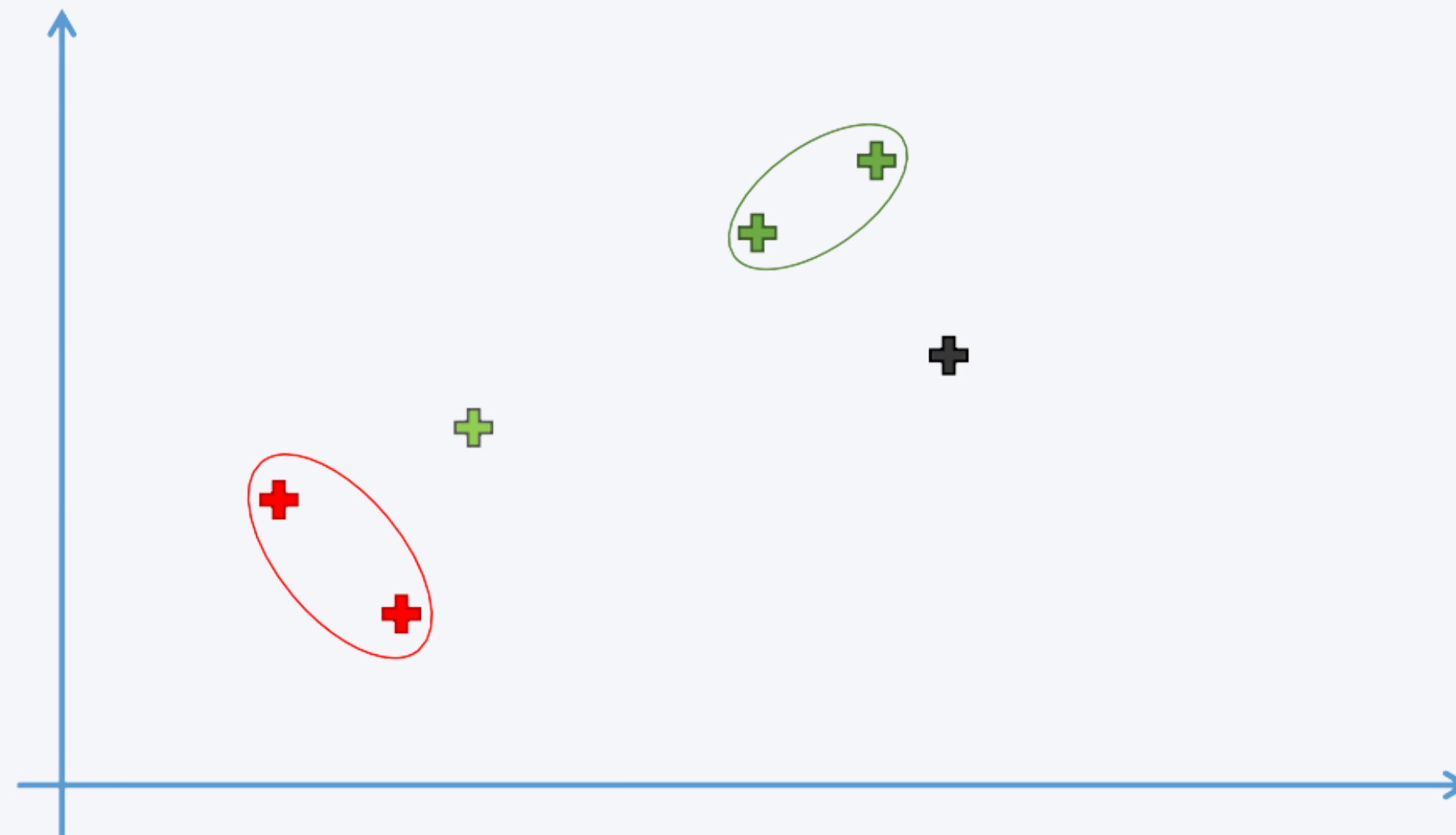
STEP 2: Take the two closest data points and make them one cluster → That forms 5 clusters



Hierarchical Clustering

Agglomerative HC

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

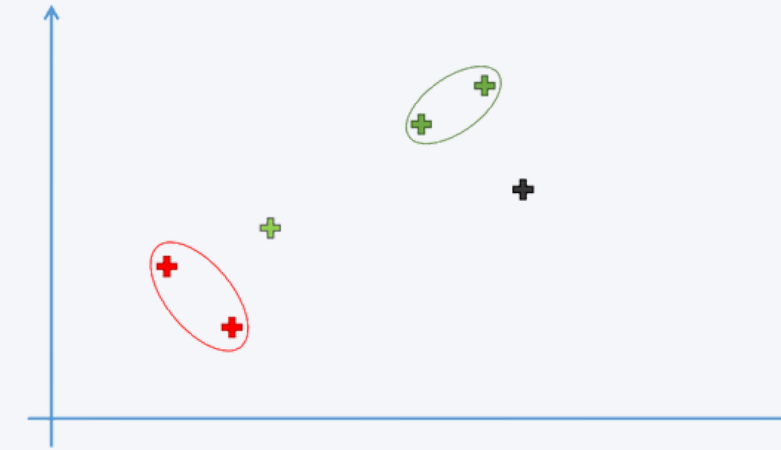


Hierarchical Clustering

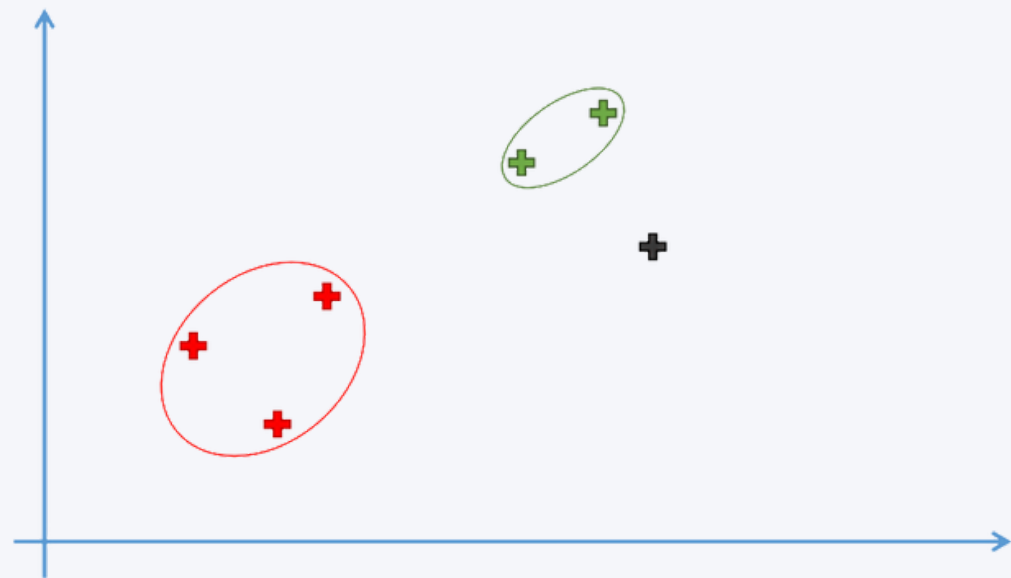


Agglomerative HC

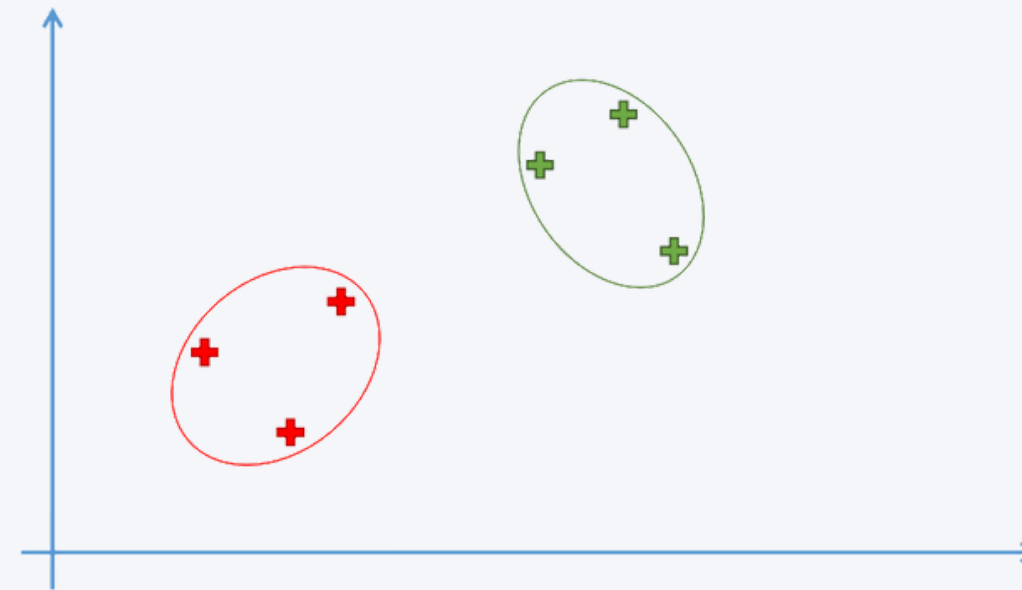
STEP 3: Take the two closest clusters and make them one cluster
→ That forms 4 clusters



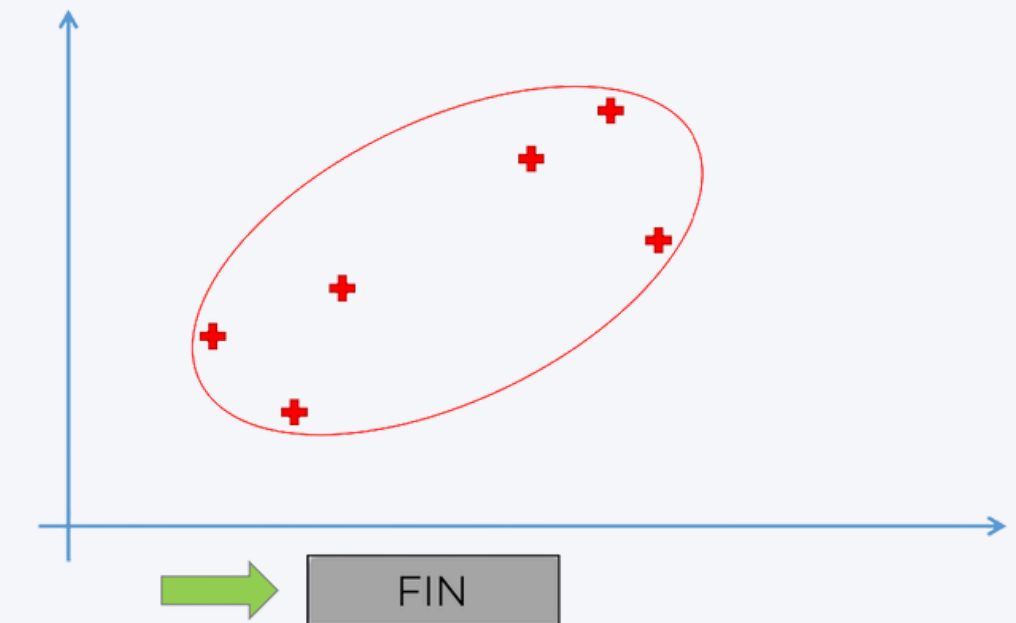
STEP 4: Repeat STEP 3 until there is only one cluster



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



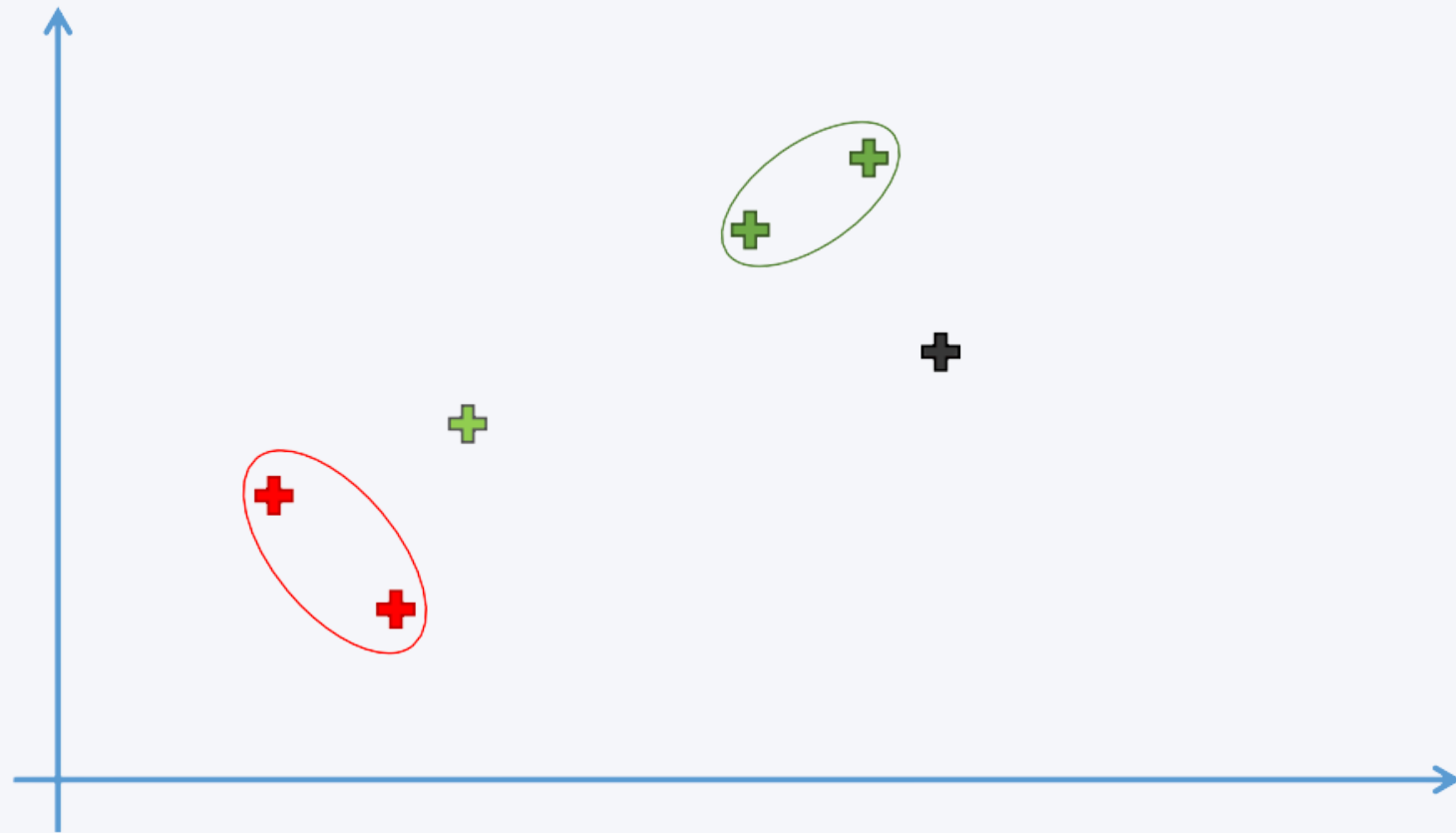
STEP 4: Repeat STEP 3 until there is only one cluster



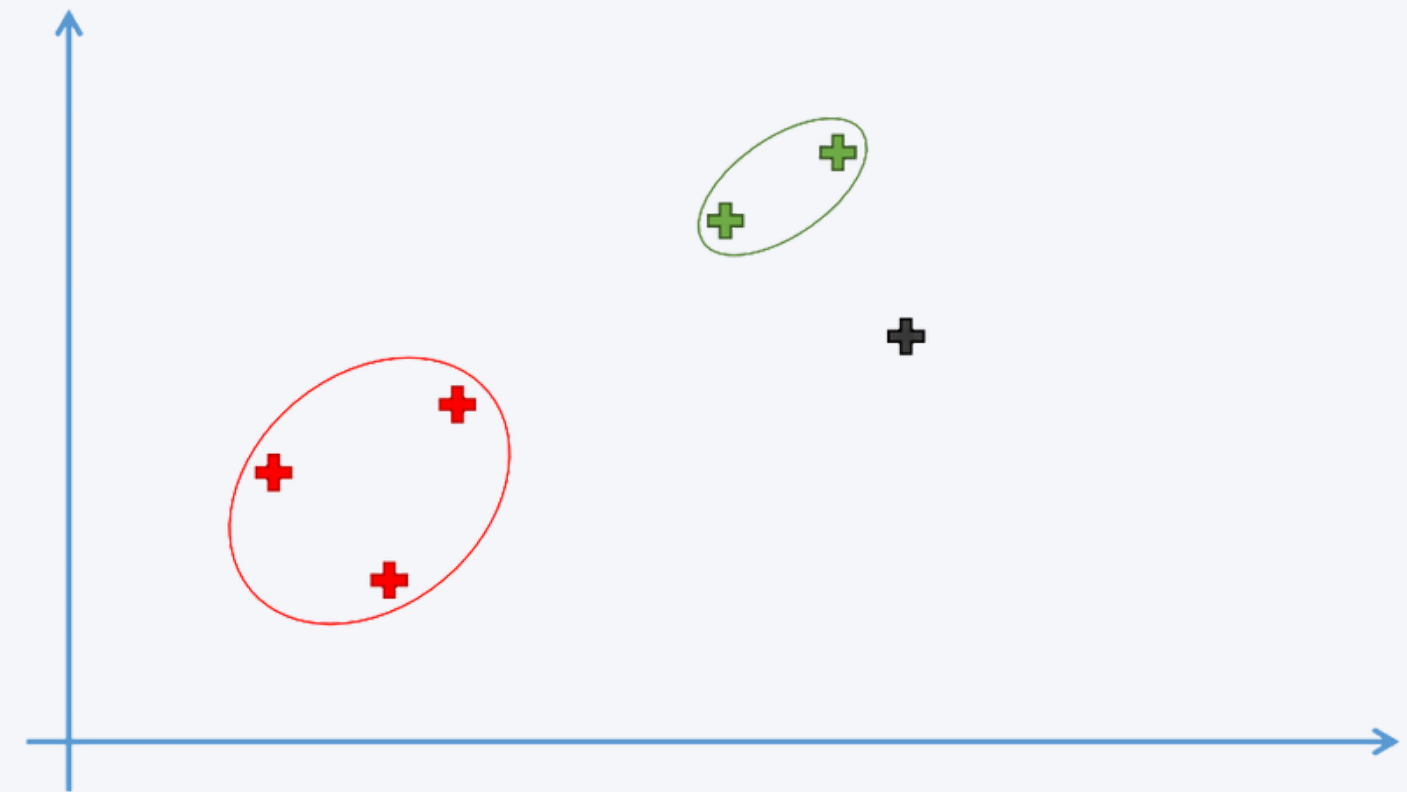
Hierarchical Clustering

Agglomerative HC

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



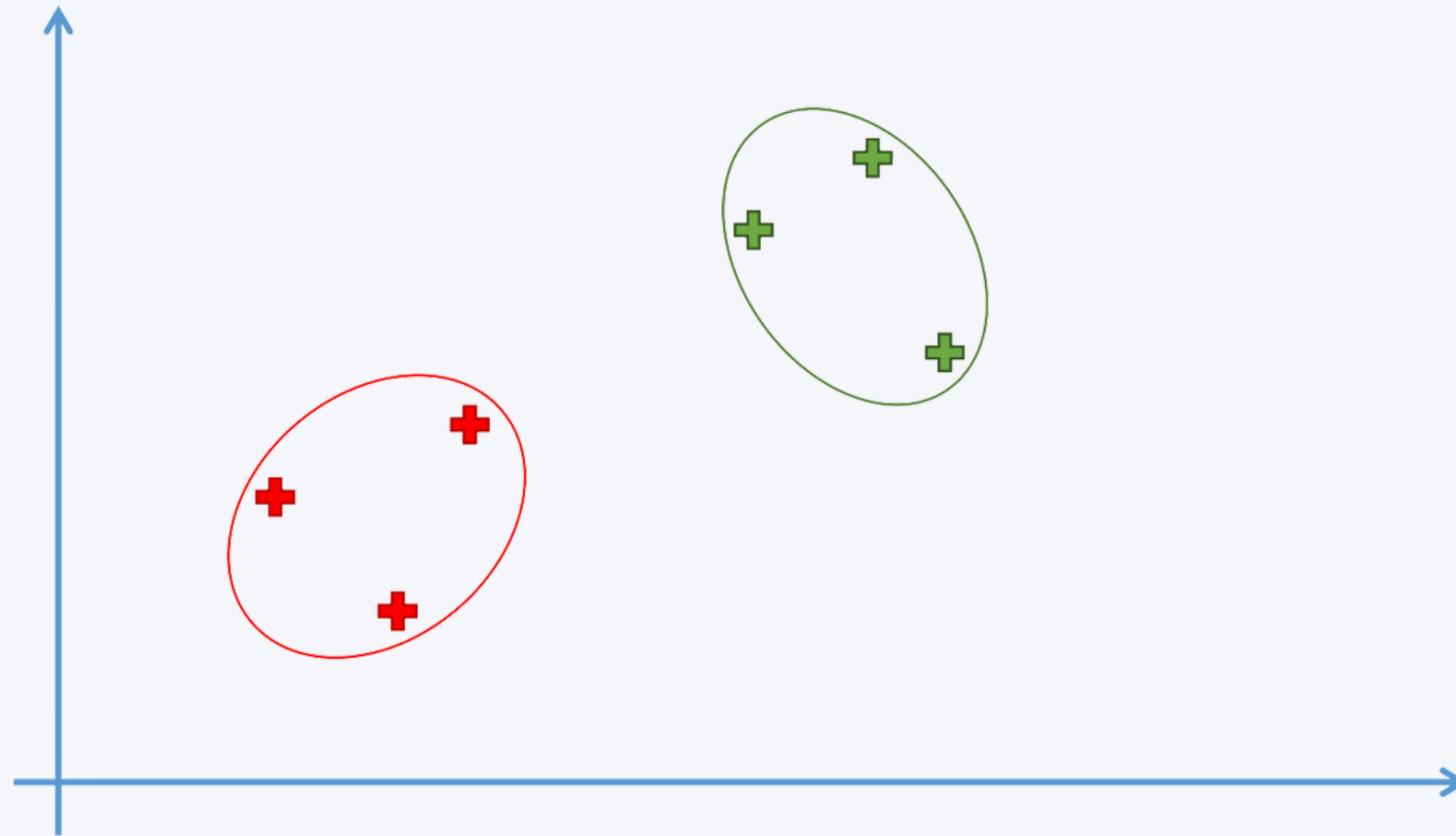
STEP 4: Repeat STEP 3 until there is only one cluster



Hierarchical Clustering

Agglomerative HC

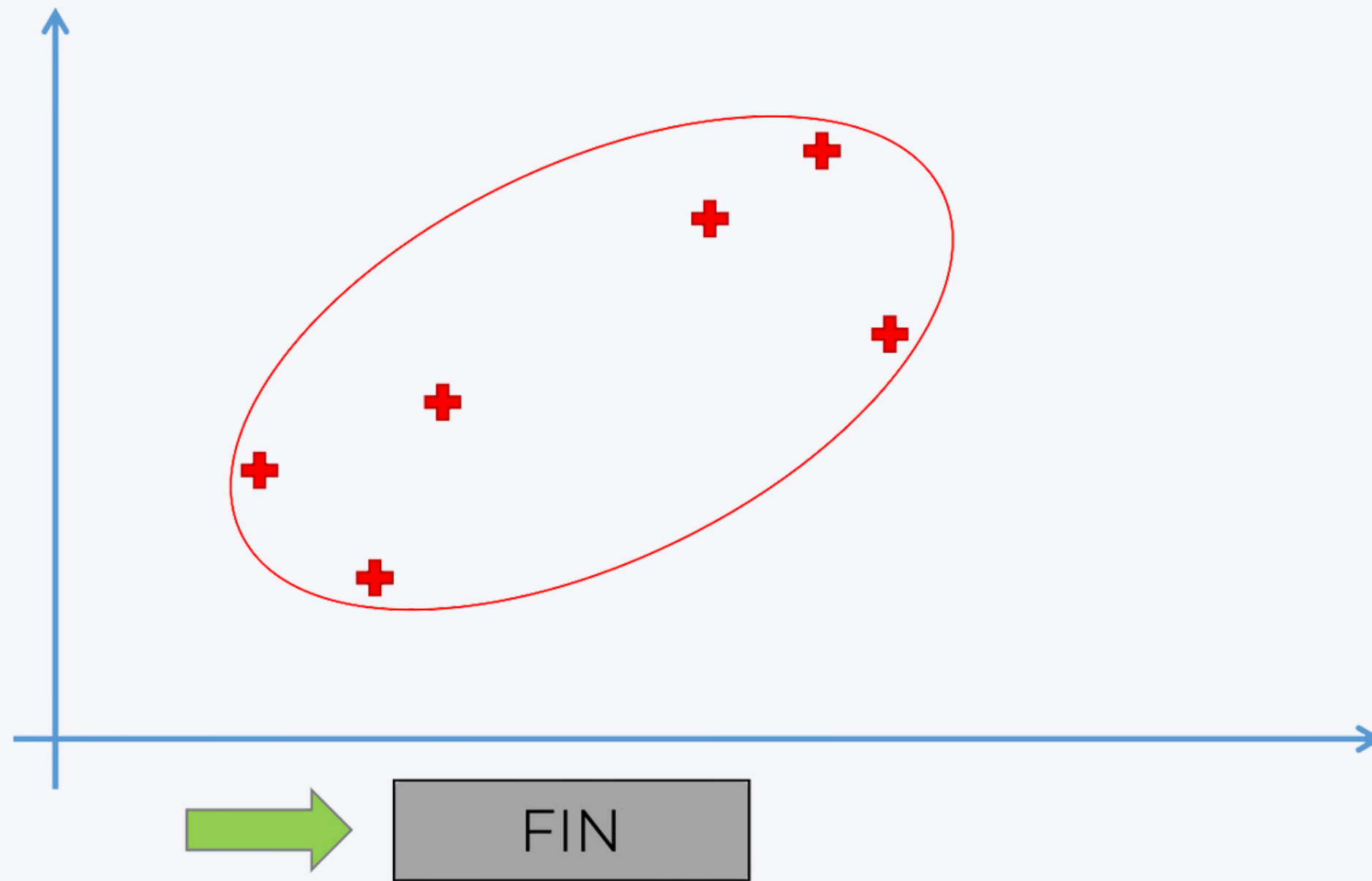
STEP 4: Repeat STEP 3 until there is only one cluster



Hierarchical Clustering

Agglomerative HC

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

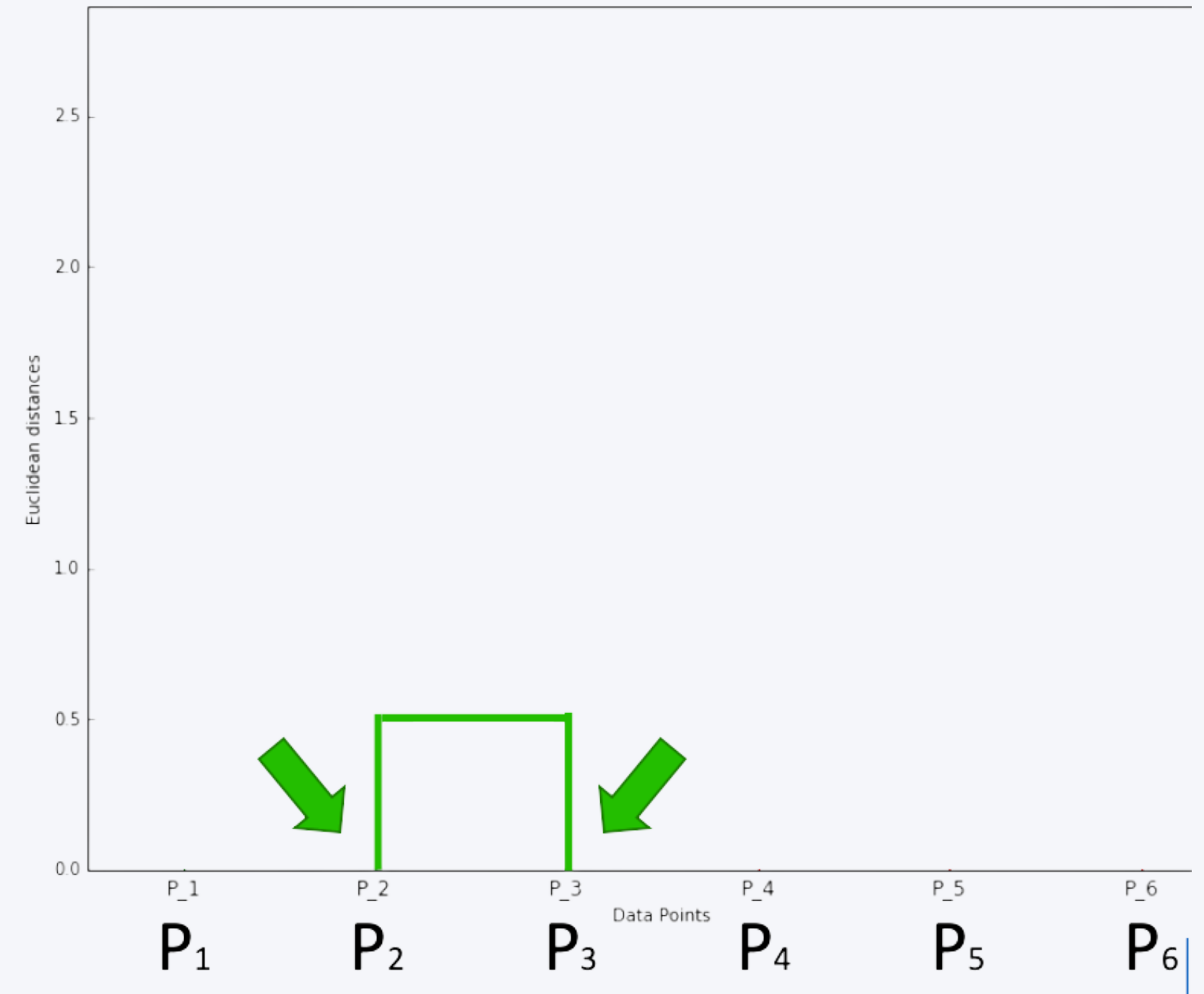
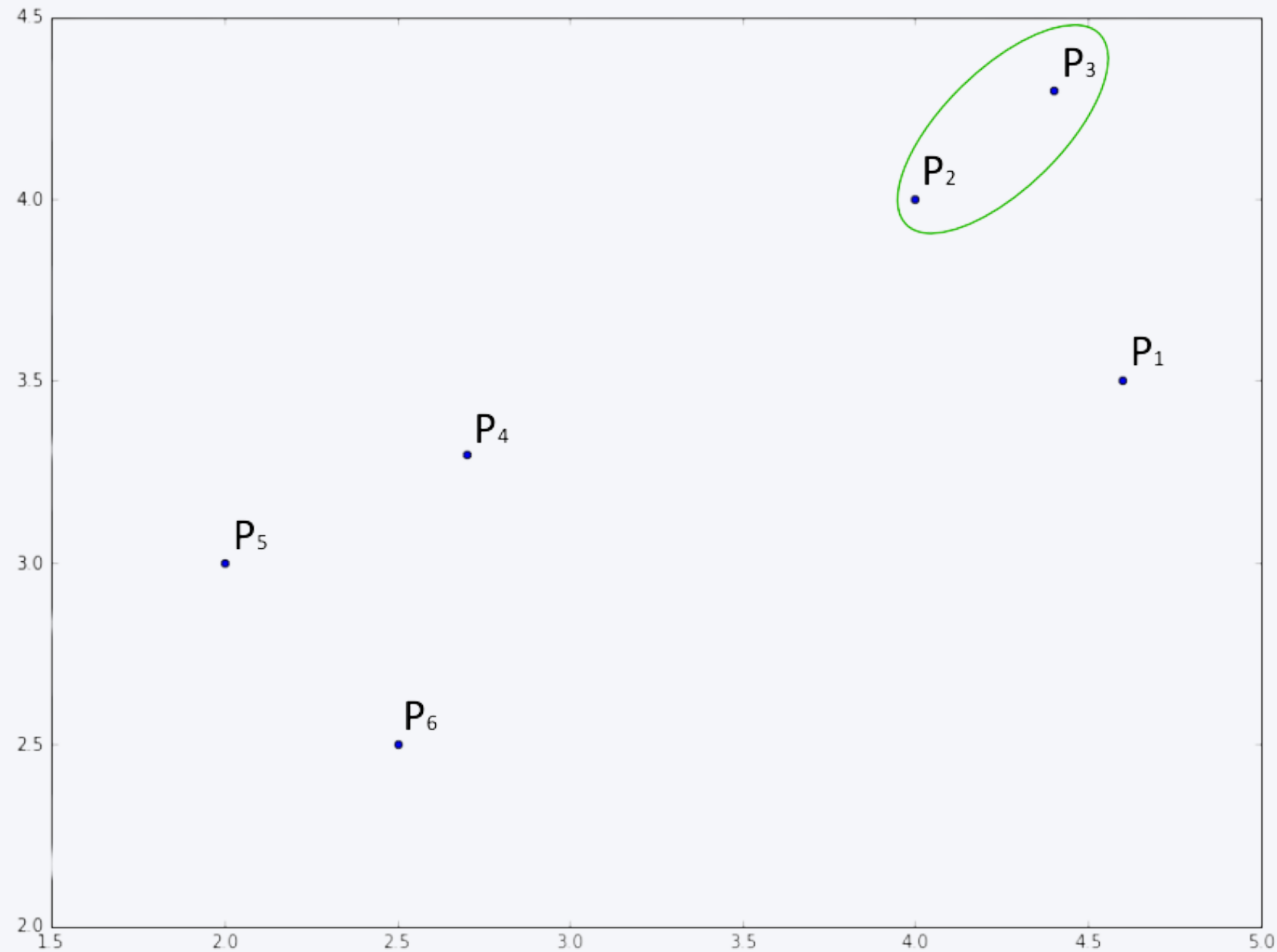


- it maintains like memory of each steps , saving the steps taken with the distances.



Hierarchical Clustering

How Do Dendograms Work ?

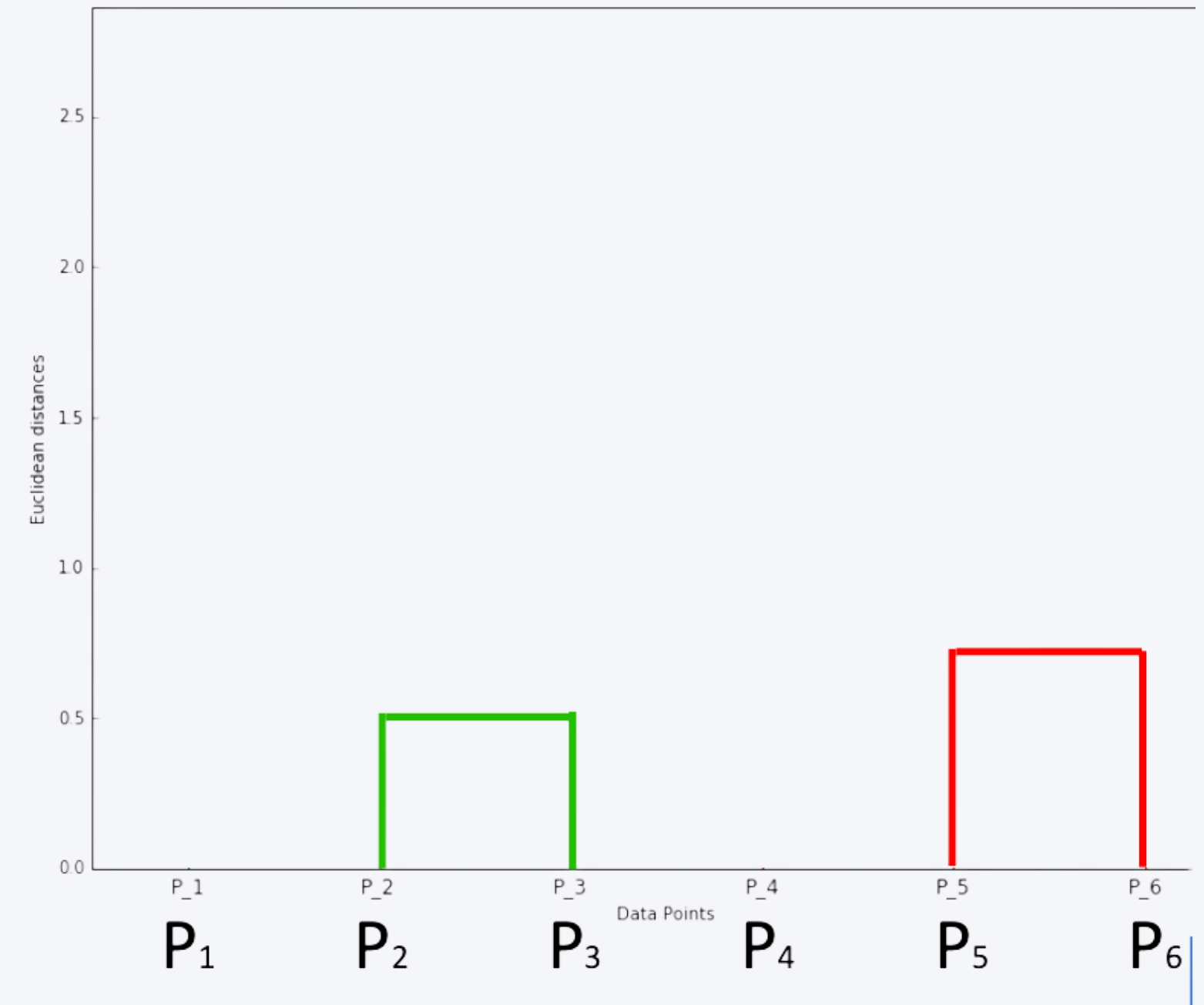
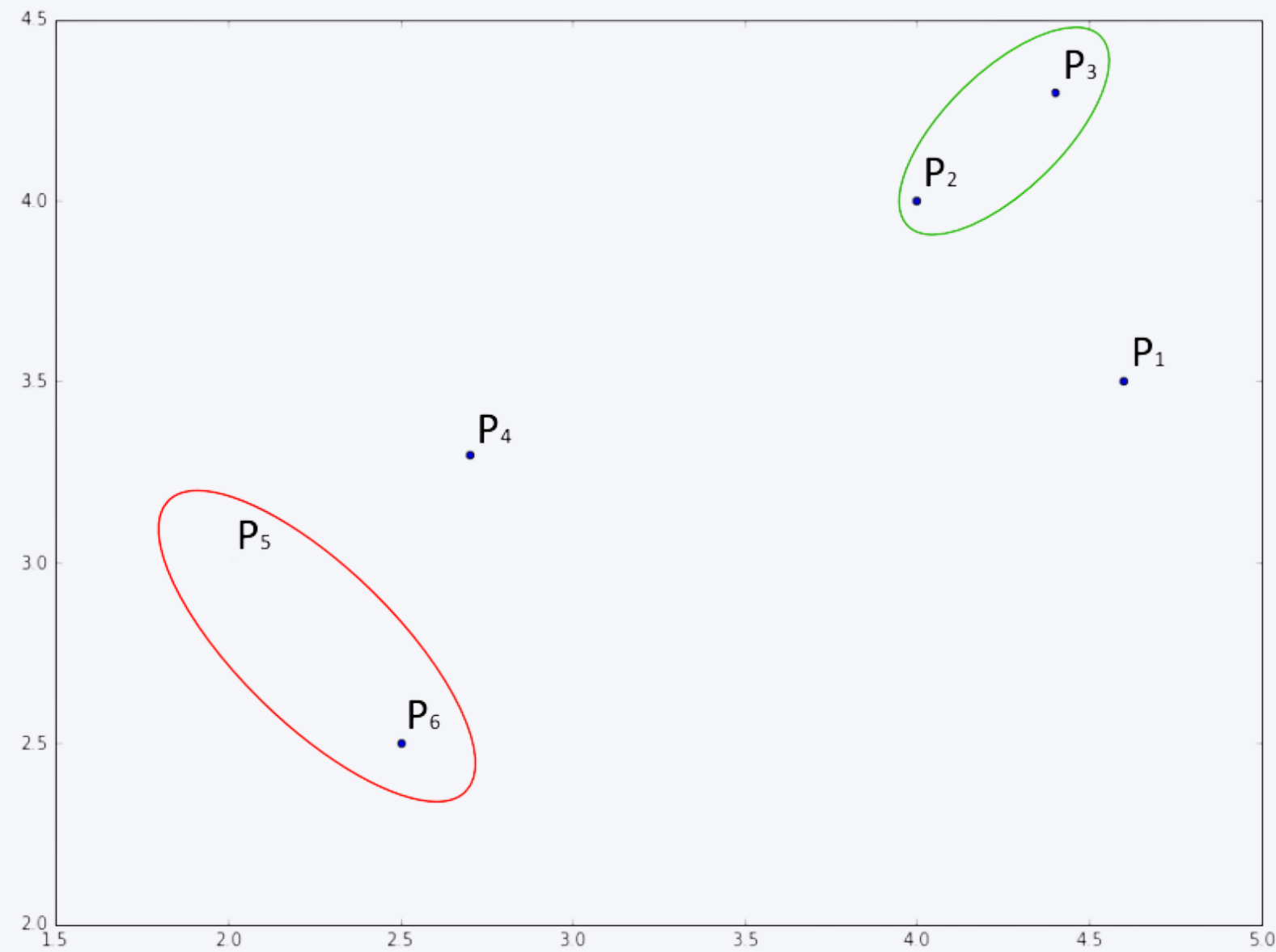


- once we defined first closest two points and made clusters, we can now assign p2 and p3 with the distance (Euclidean), and it is the similarity of these clusters



Hierarchical Clustering

How Do Dendograms Work ?

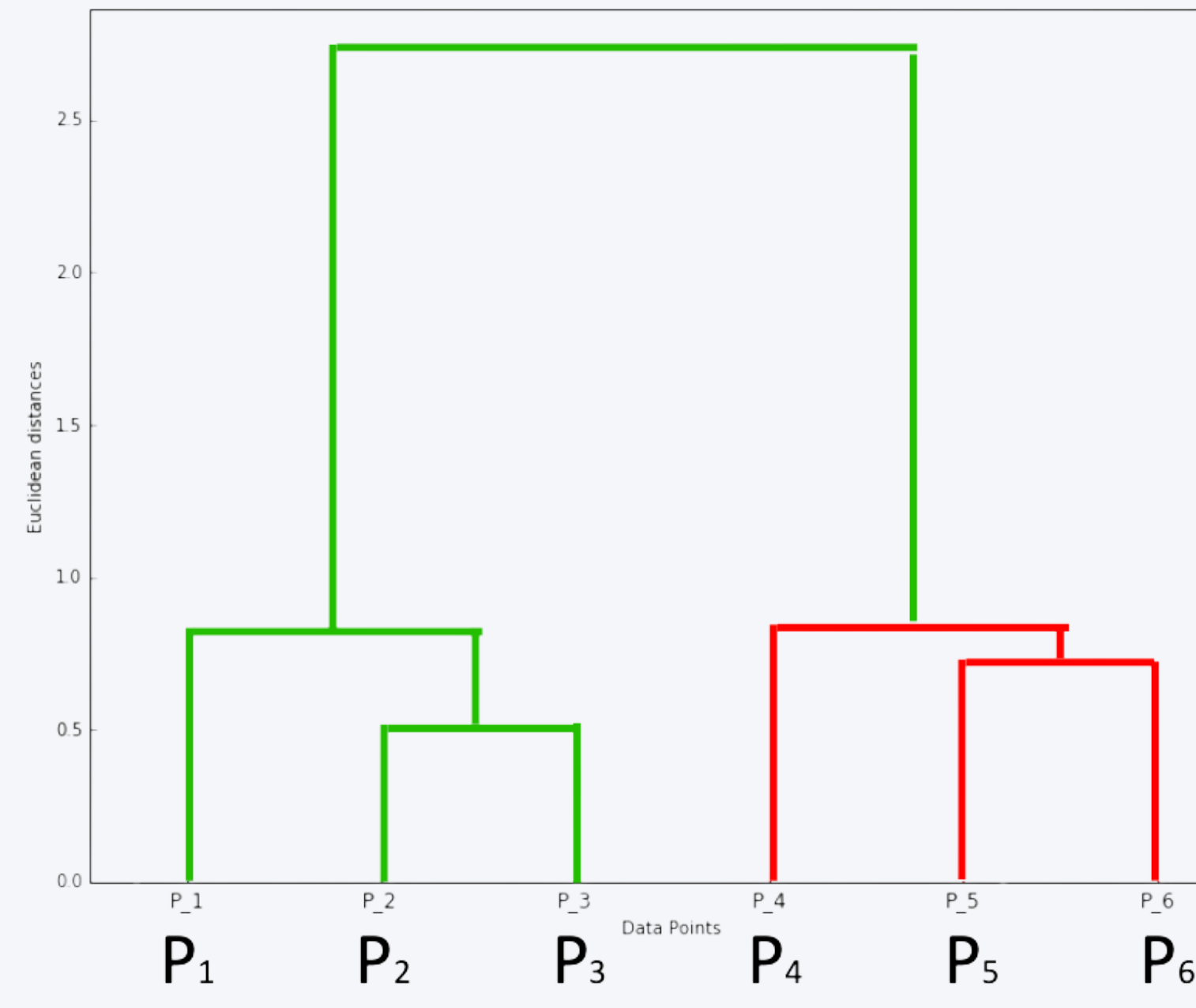
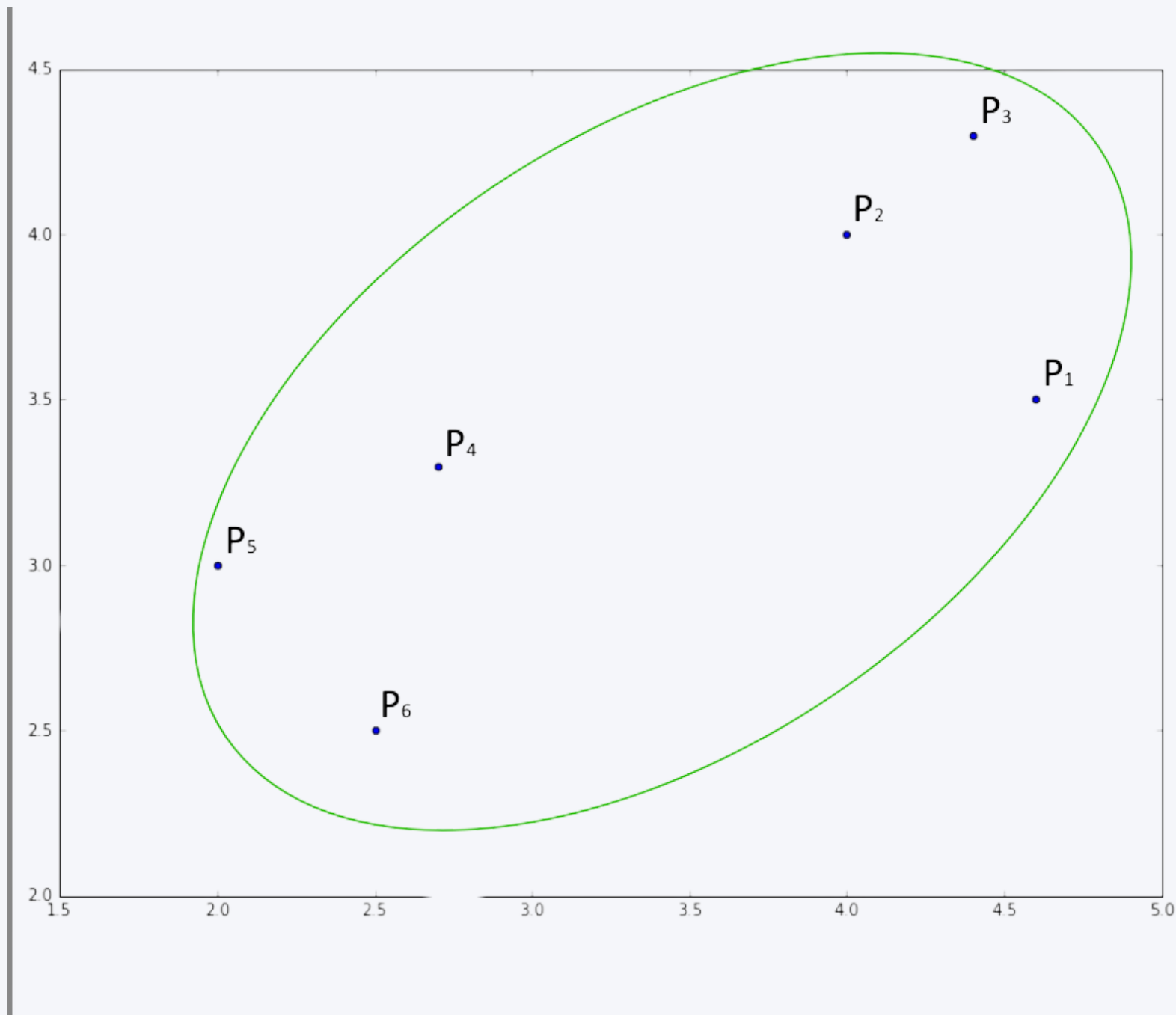


- Next step , the next two closest clusters and plot them as Dendograms

Hierarchical Clustering



How Do Dendograms Work ?

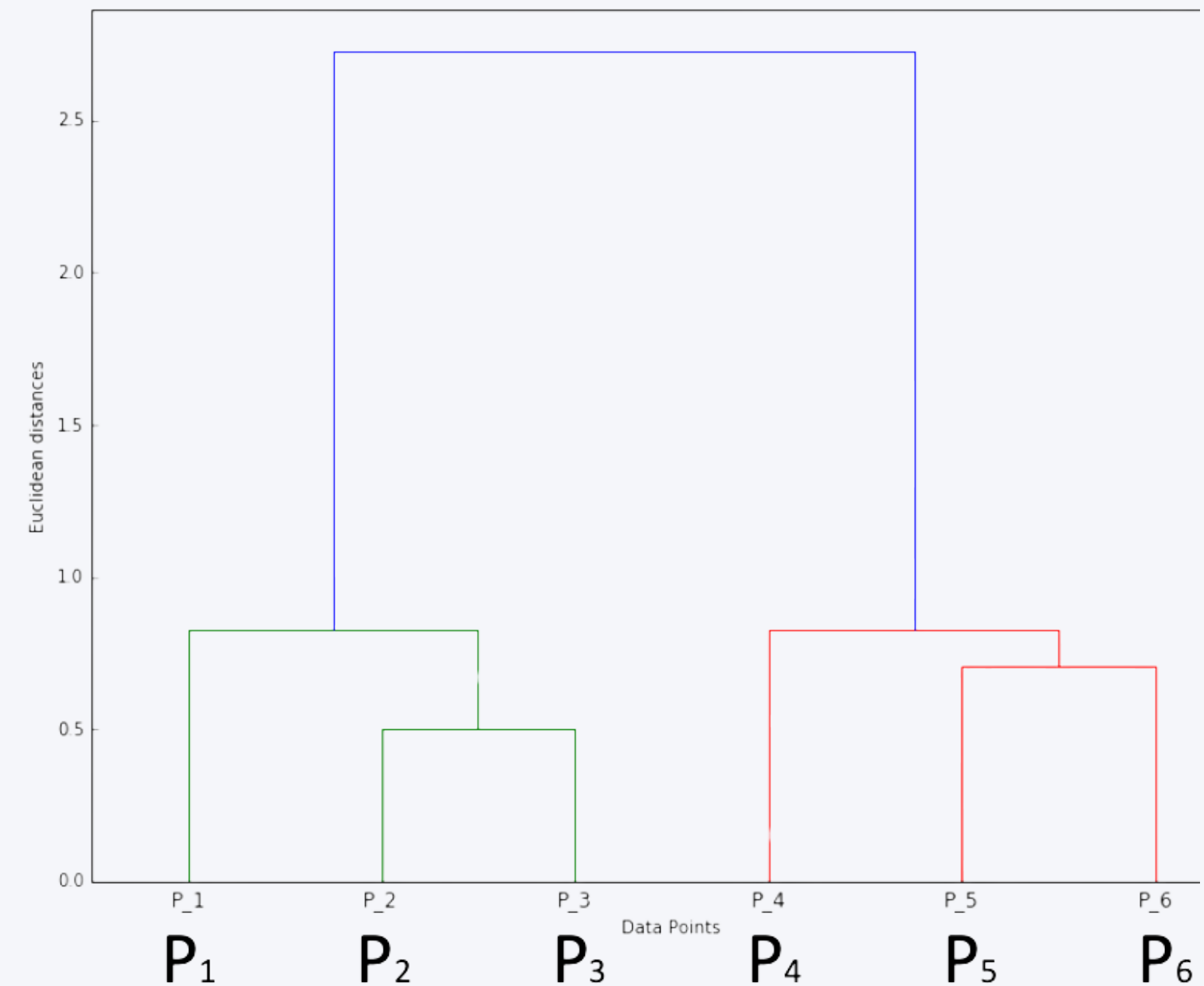
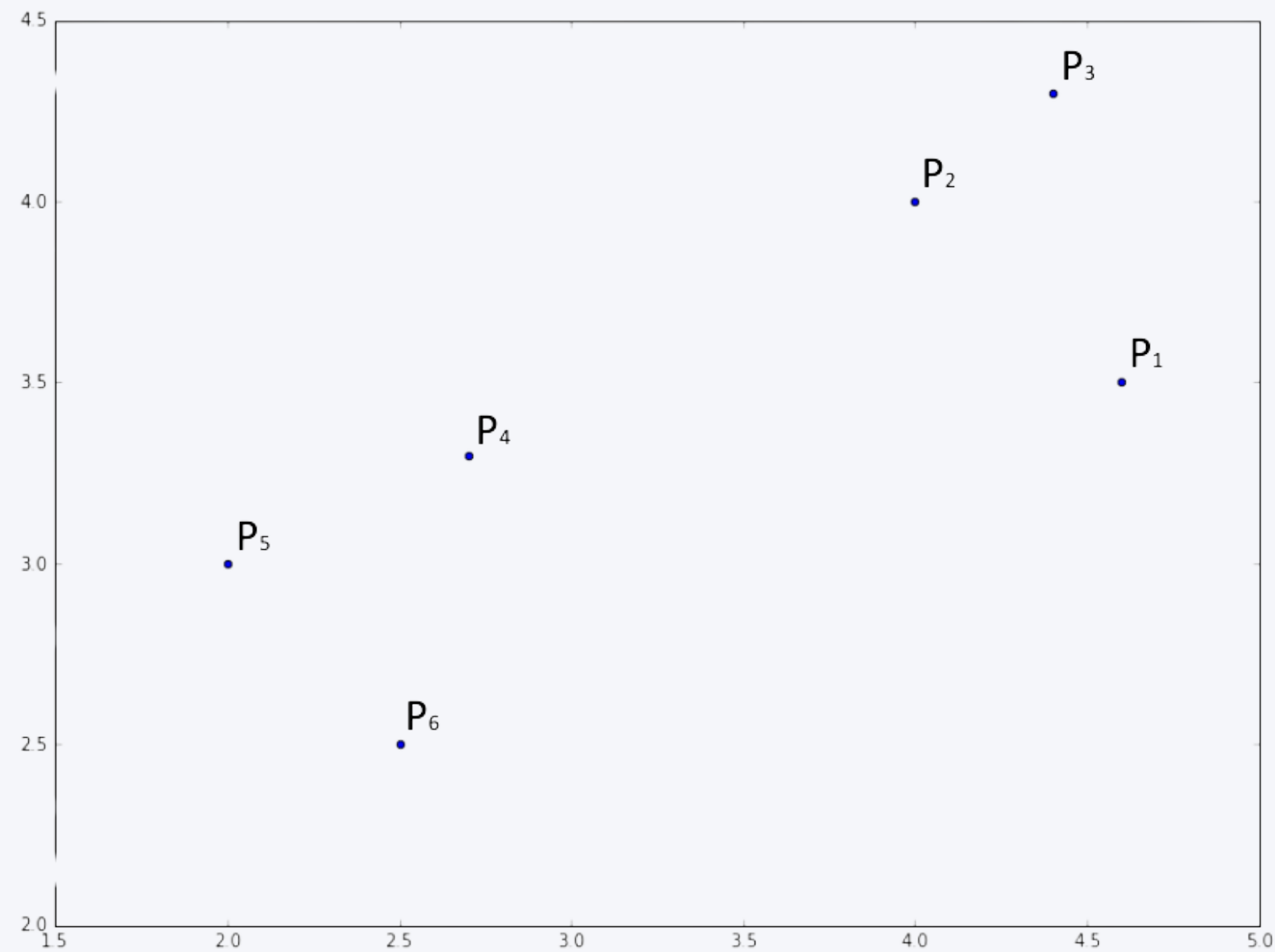


- Repeat the process until you get one cluster

Hierarchical Clustering



How Do Dendograms Work ?

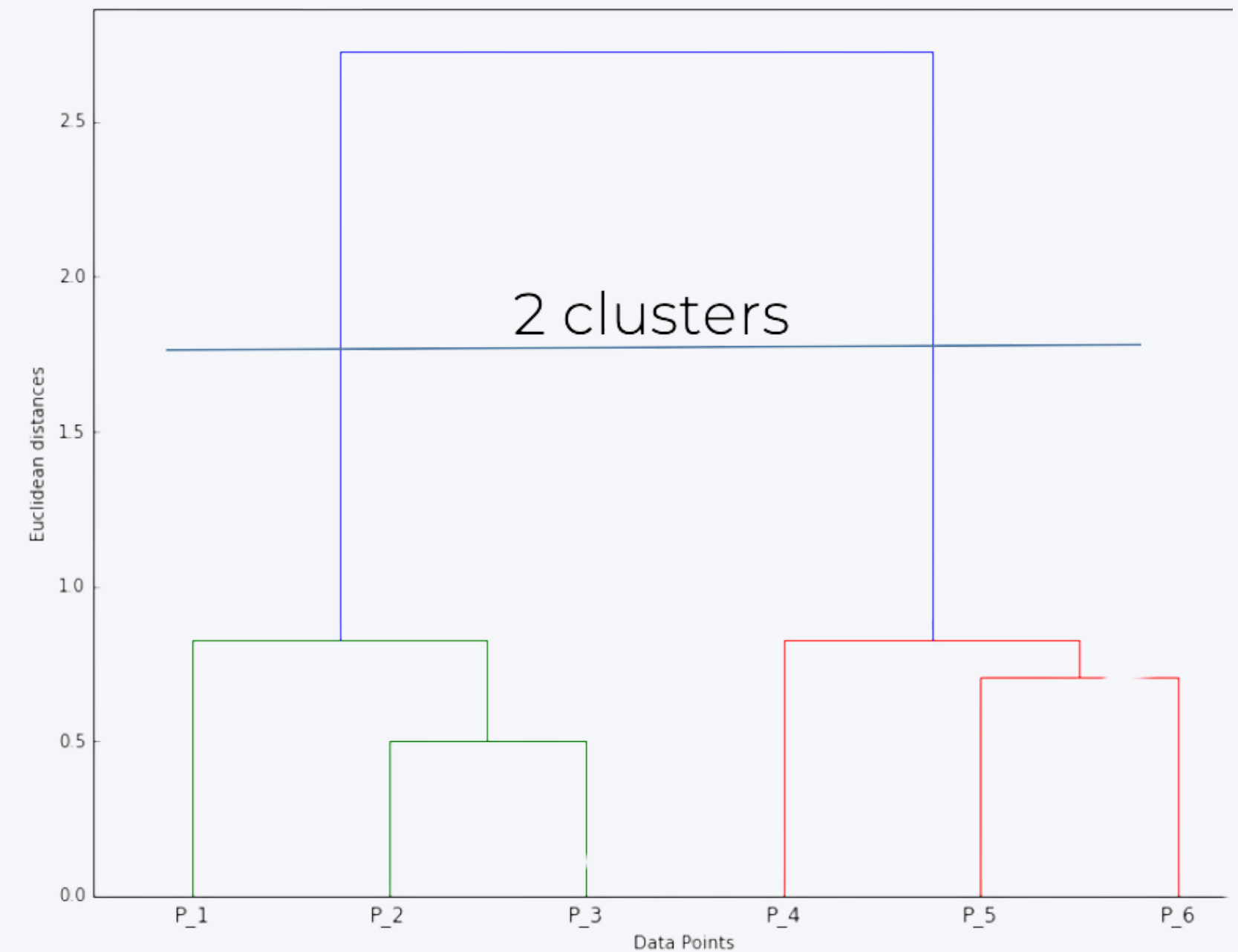
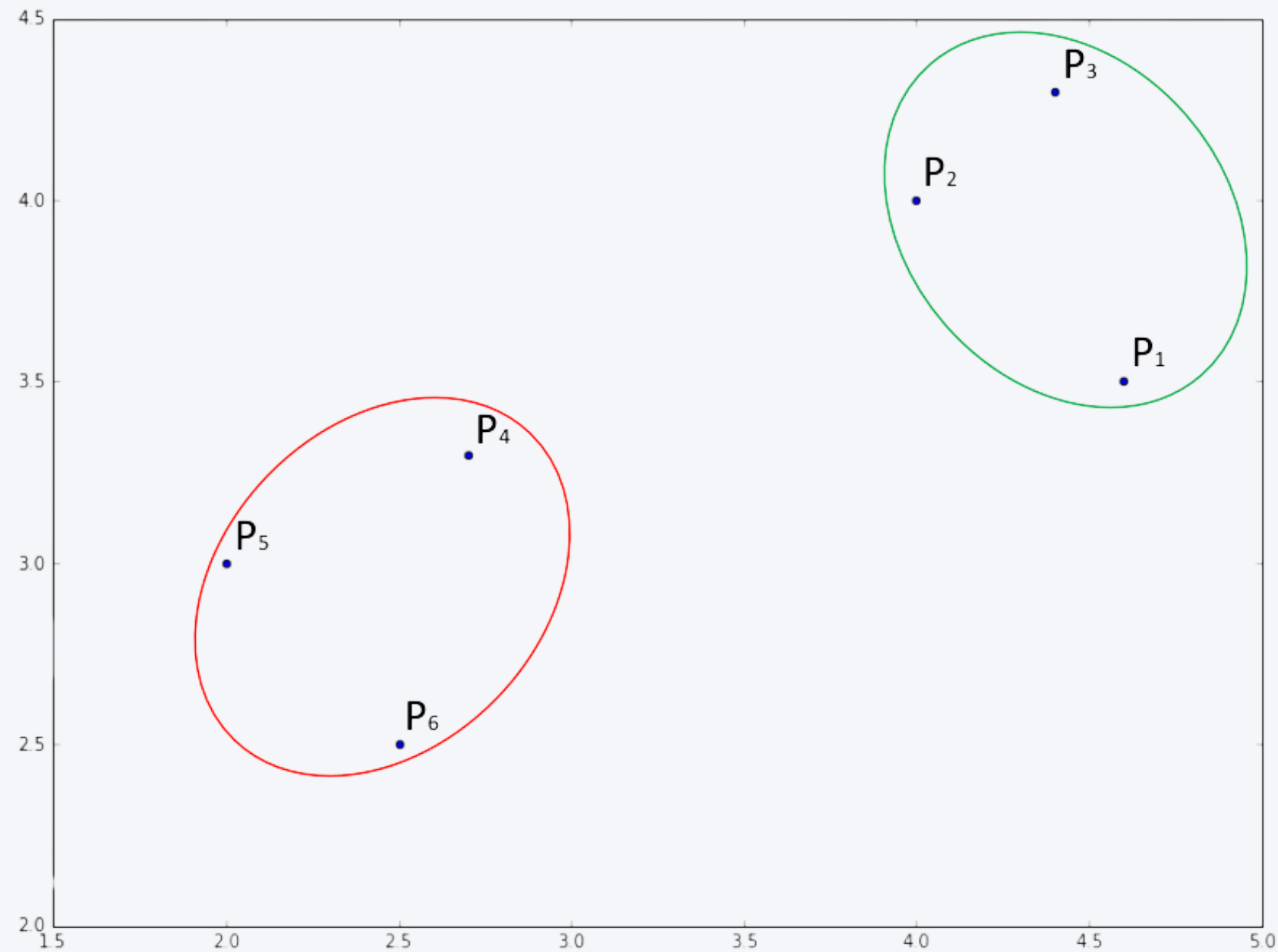


- Final Dendograms with clusters and distances (Memory on how clusters are formed)

Hierarchical Clustering



Using Dendograms ?

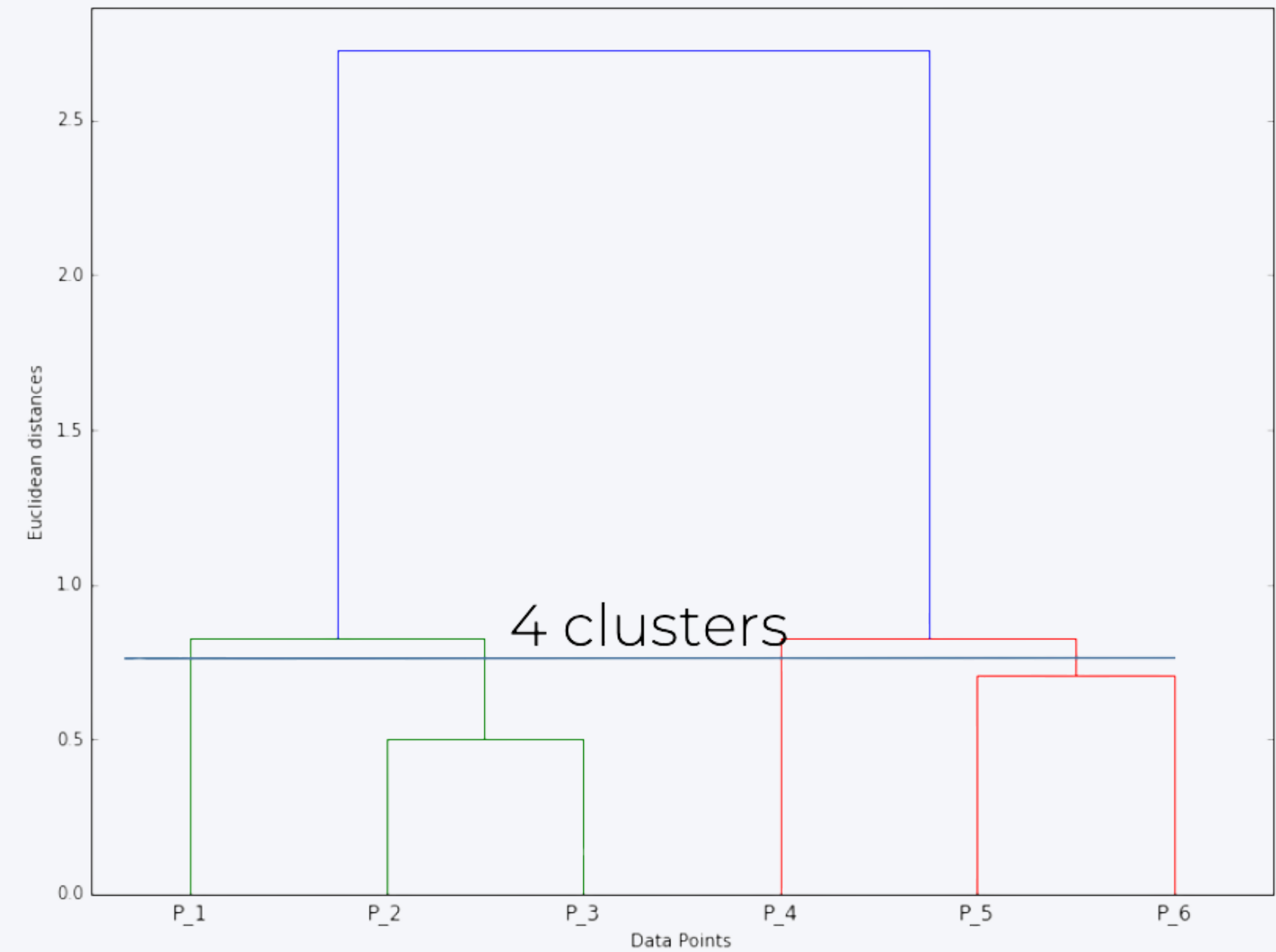
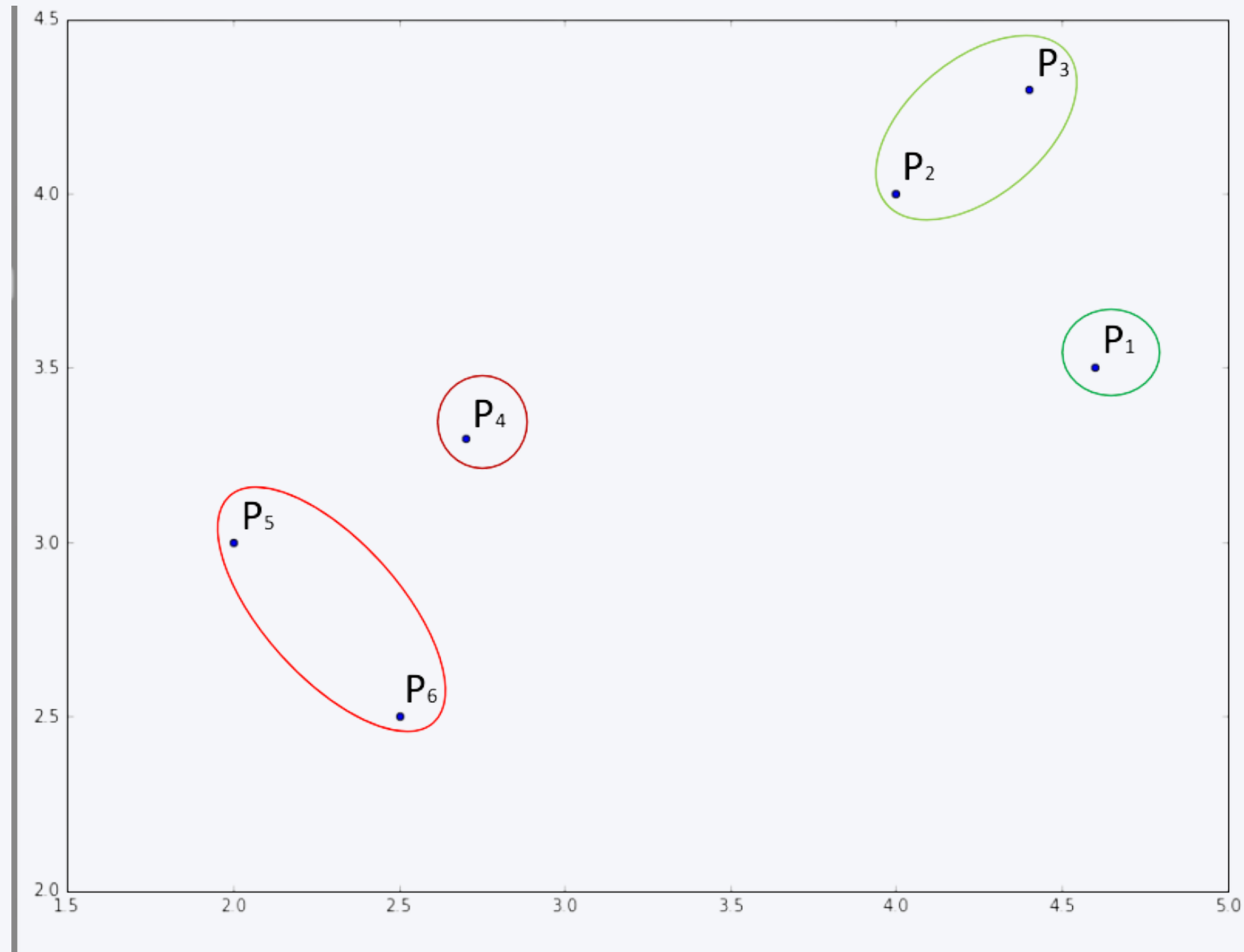


- We can set threshold (distance threshold or dissimilarities) which help in making better clusters by increasing similarity of these clusters // by vertical lines, we can see how many clusters we have.

Hierarchical Clustering



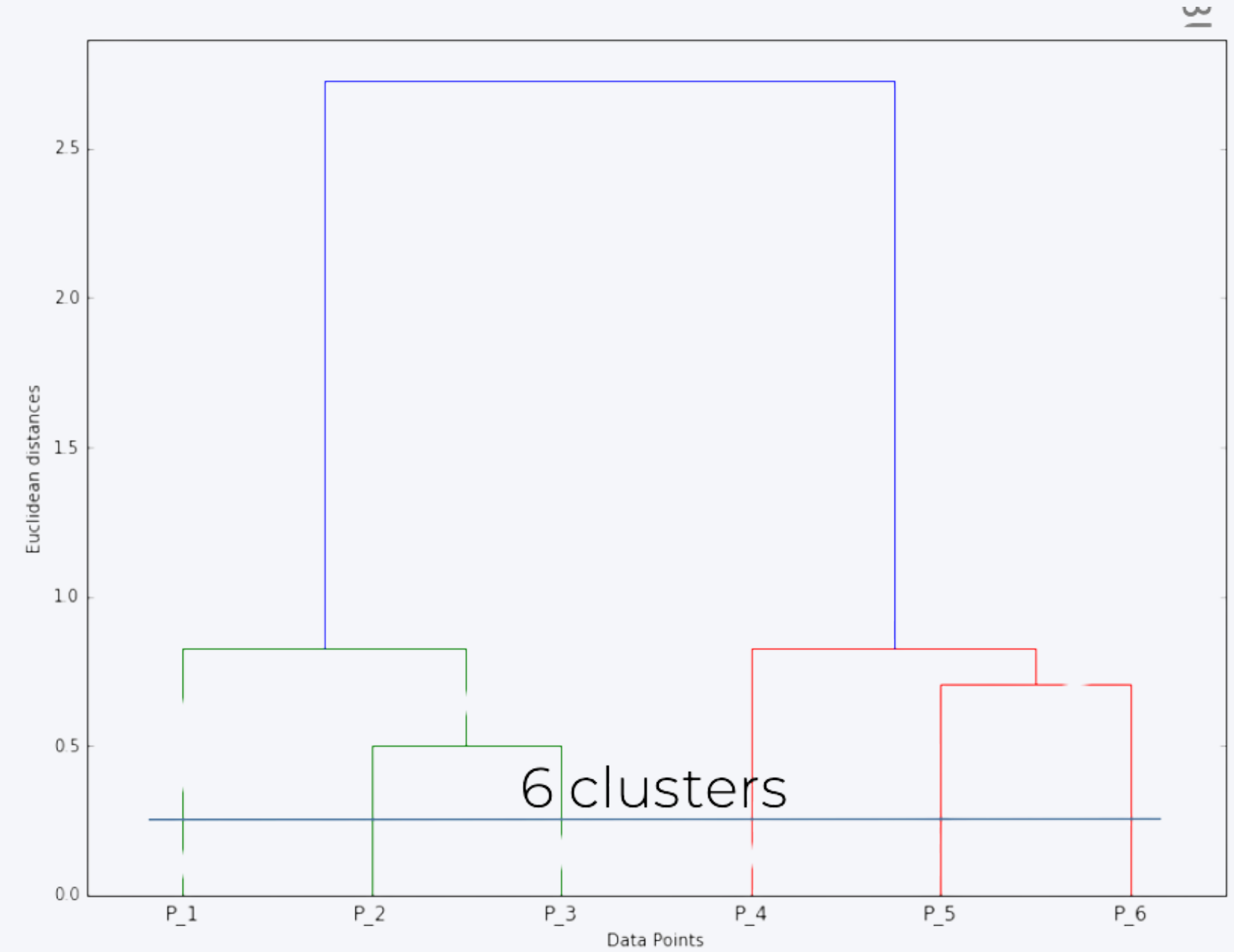
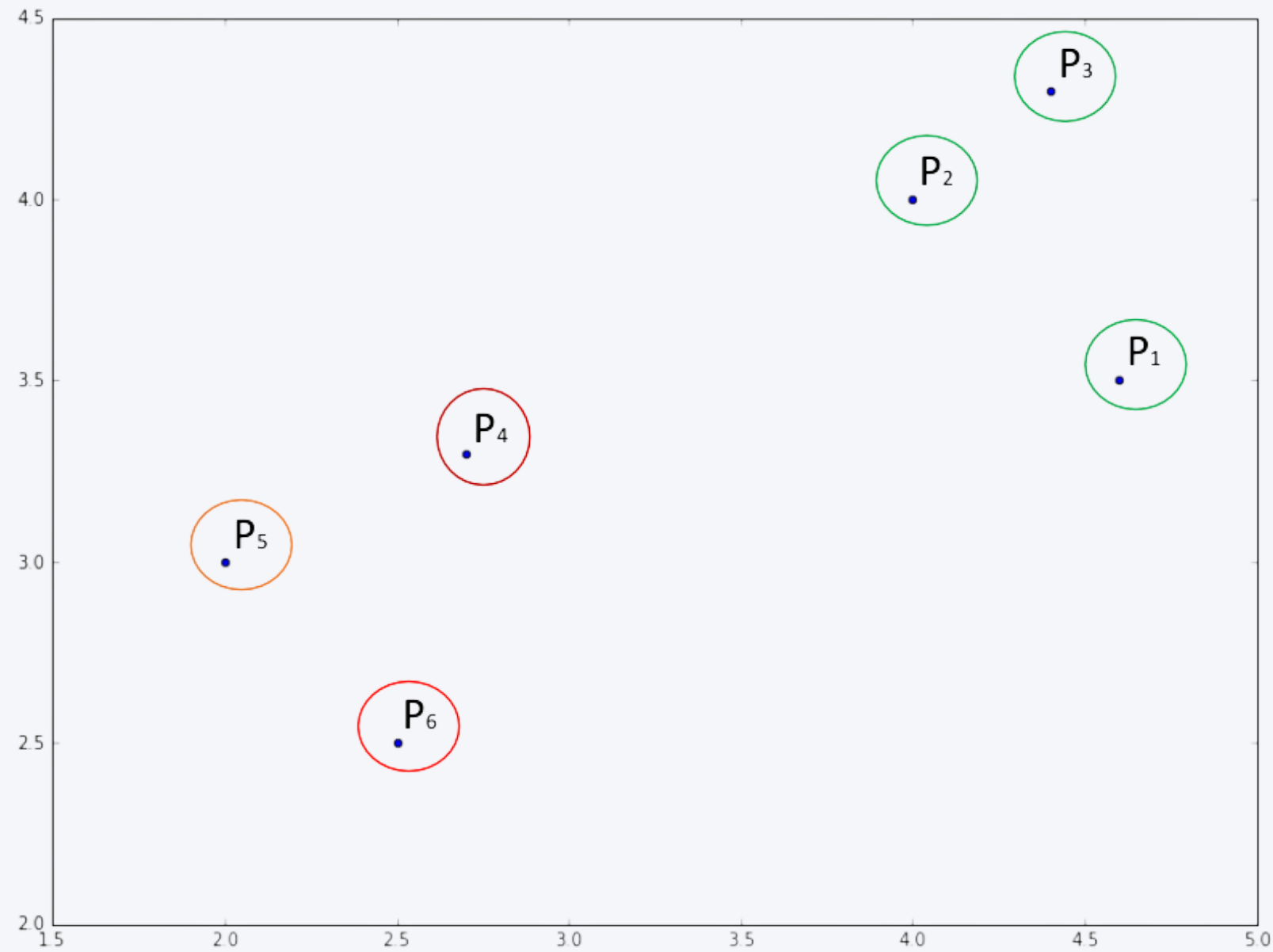
Using Dendograms ?



- Another threshold

Hierarchical Clustering

Using Dendograms ?

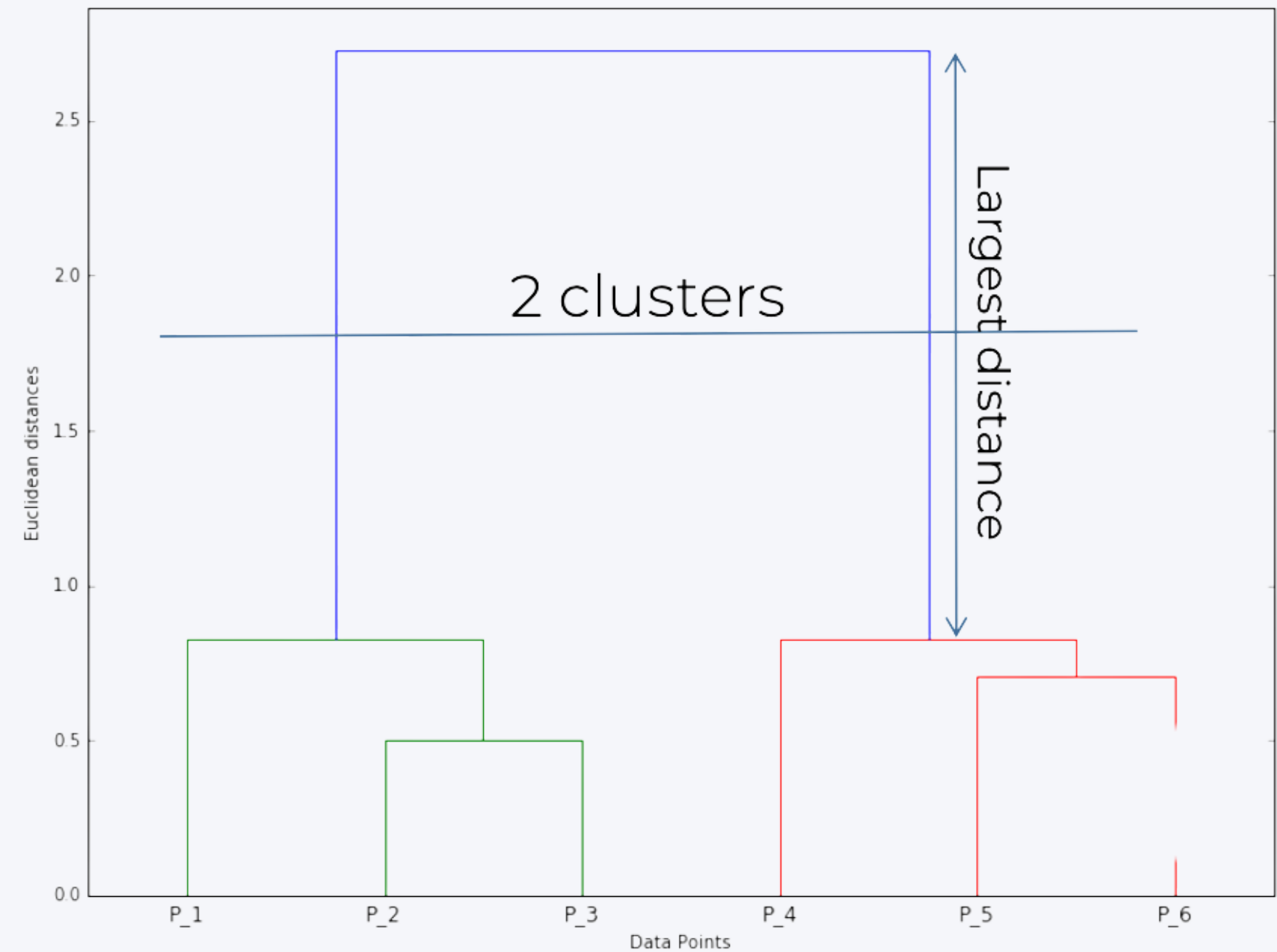
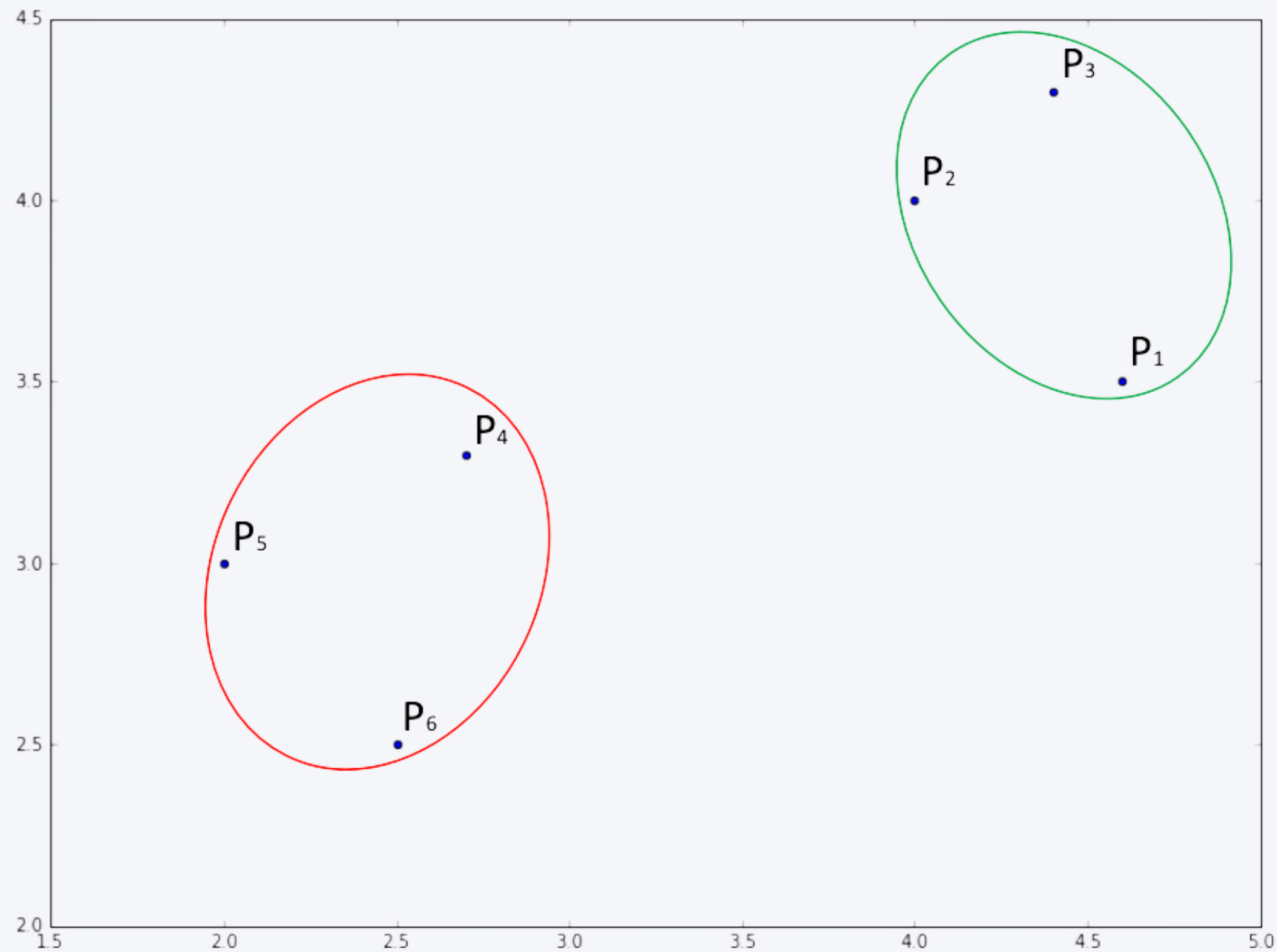


- Threshold below our first clusters

Hierarchical Clustering



Using Dendograms | Optimal Number of clusters:



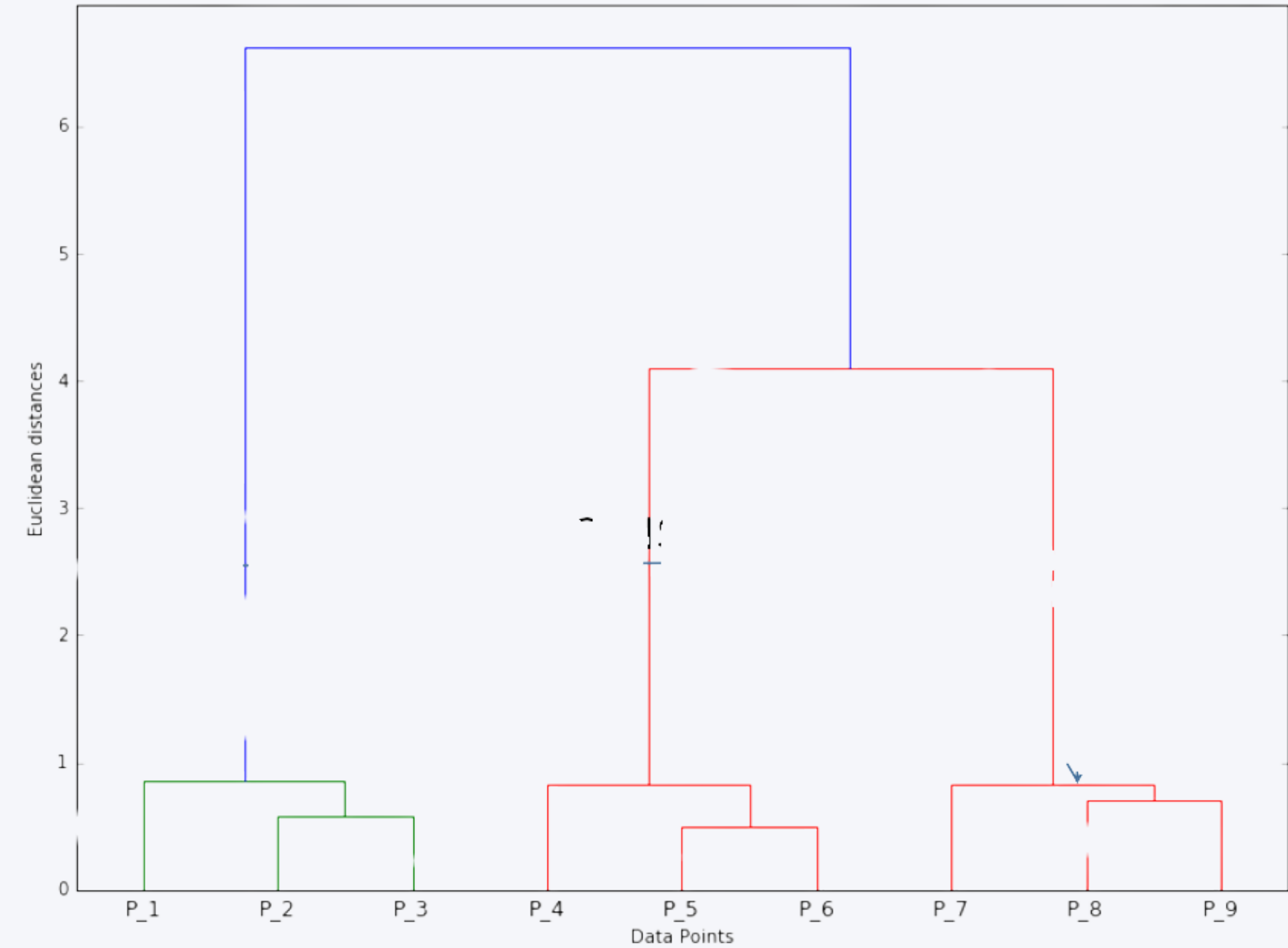
- By using vertical distance that **not** cross any horizontal line and should be **longest** line
-

Hierarchical Clustering

Dendograms | Knowledge Test



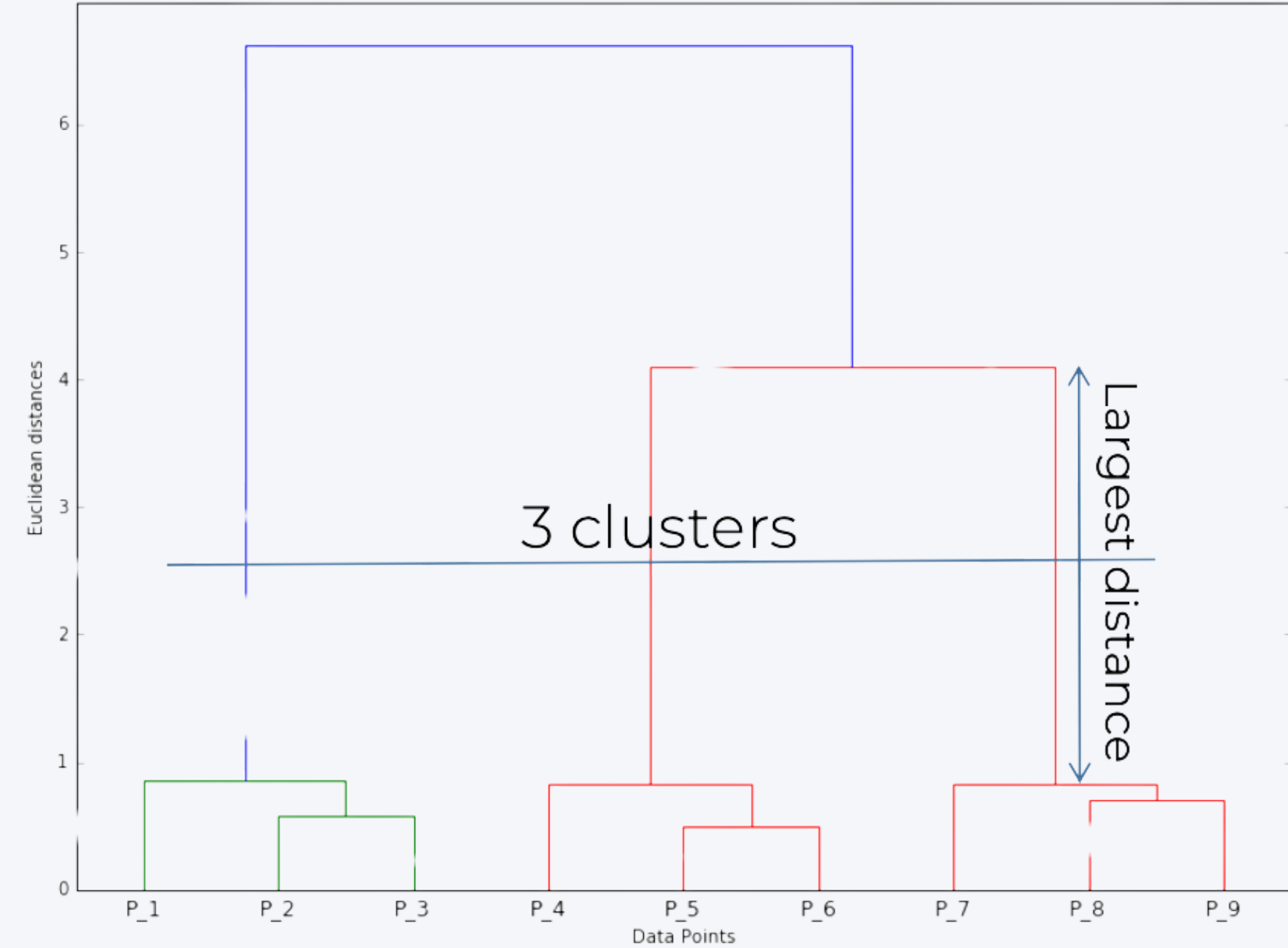
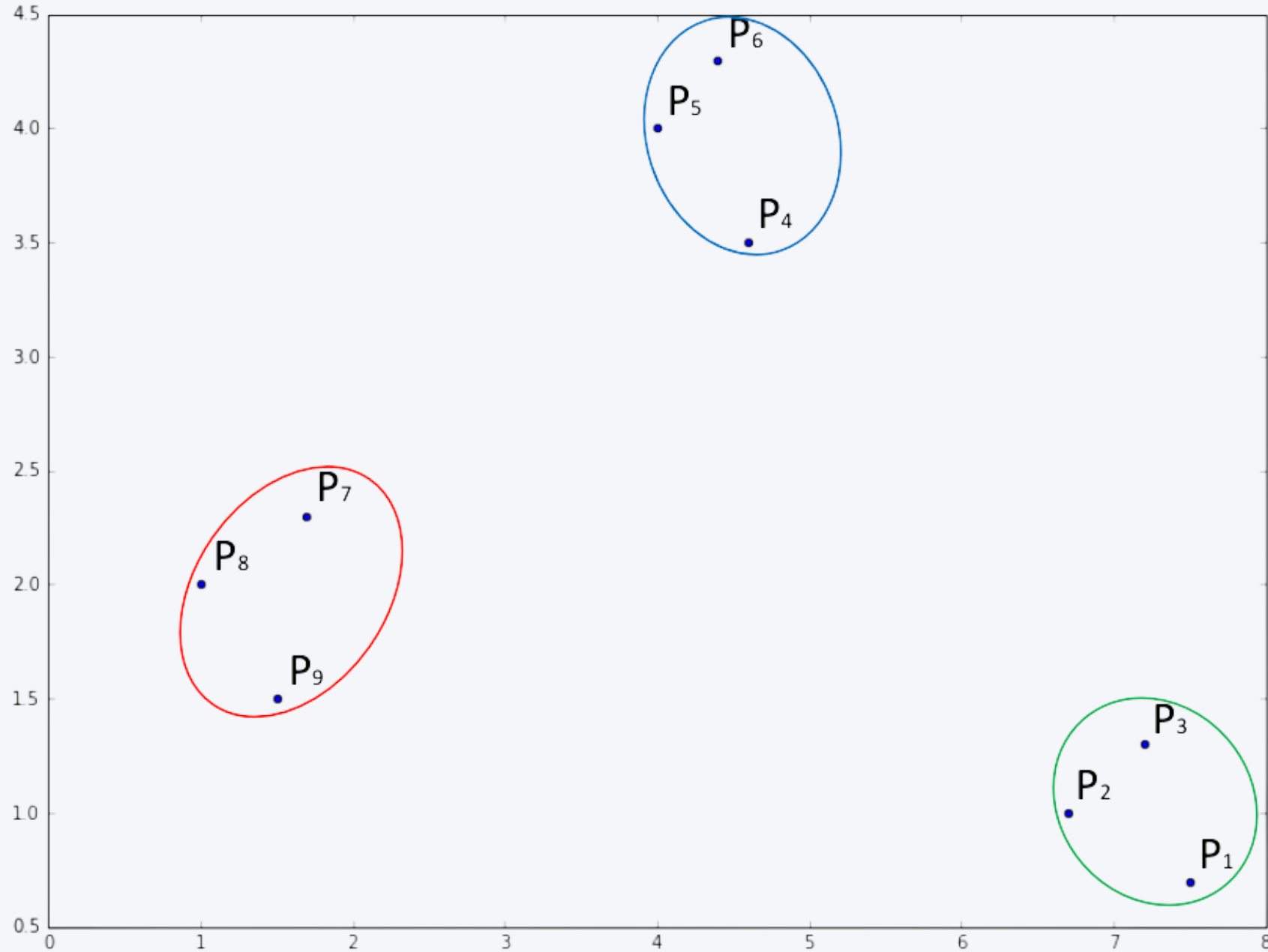
???





Hierarchical Clustering

Dendograms | Knowledge Test





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