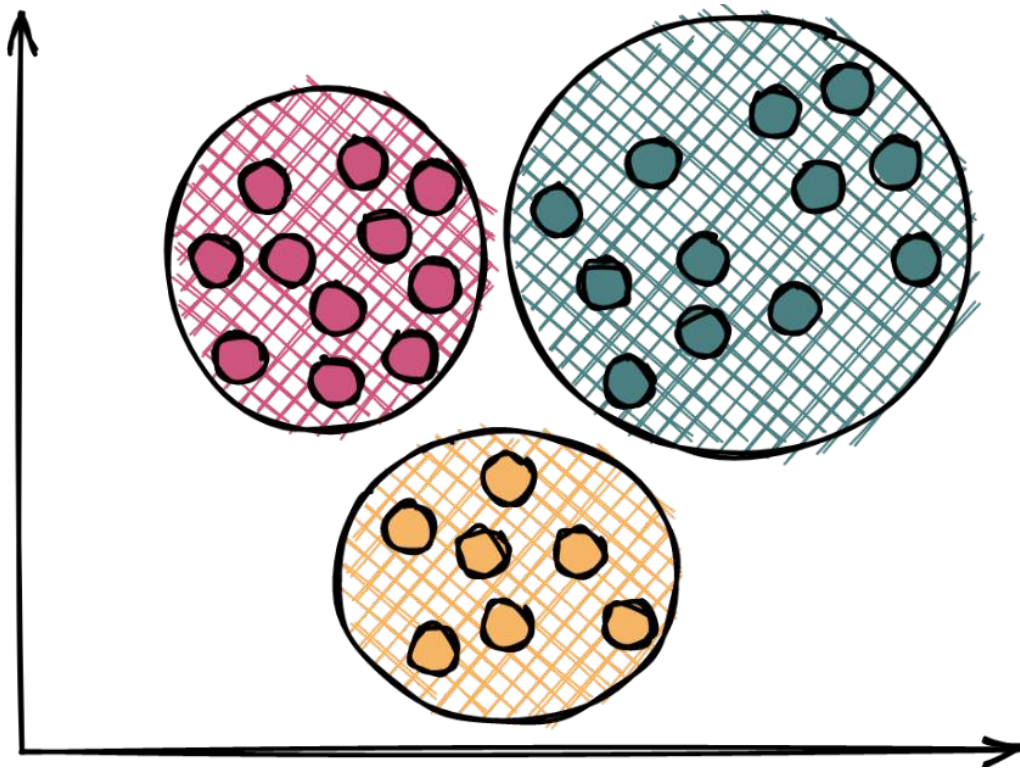


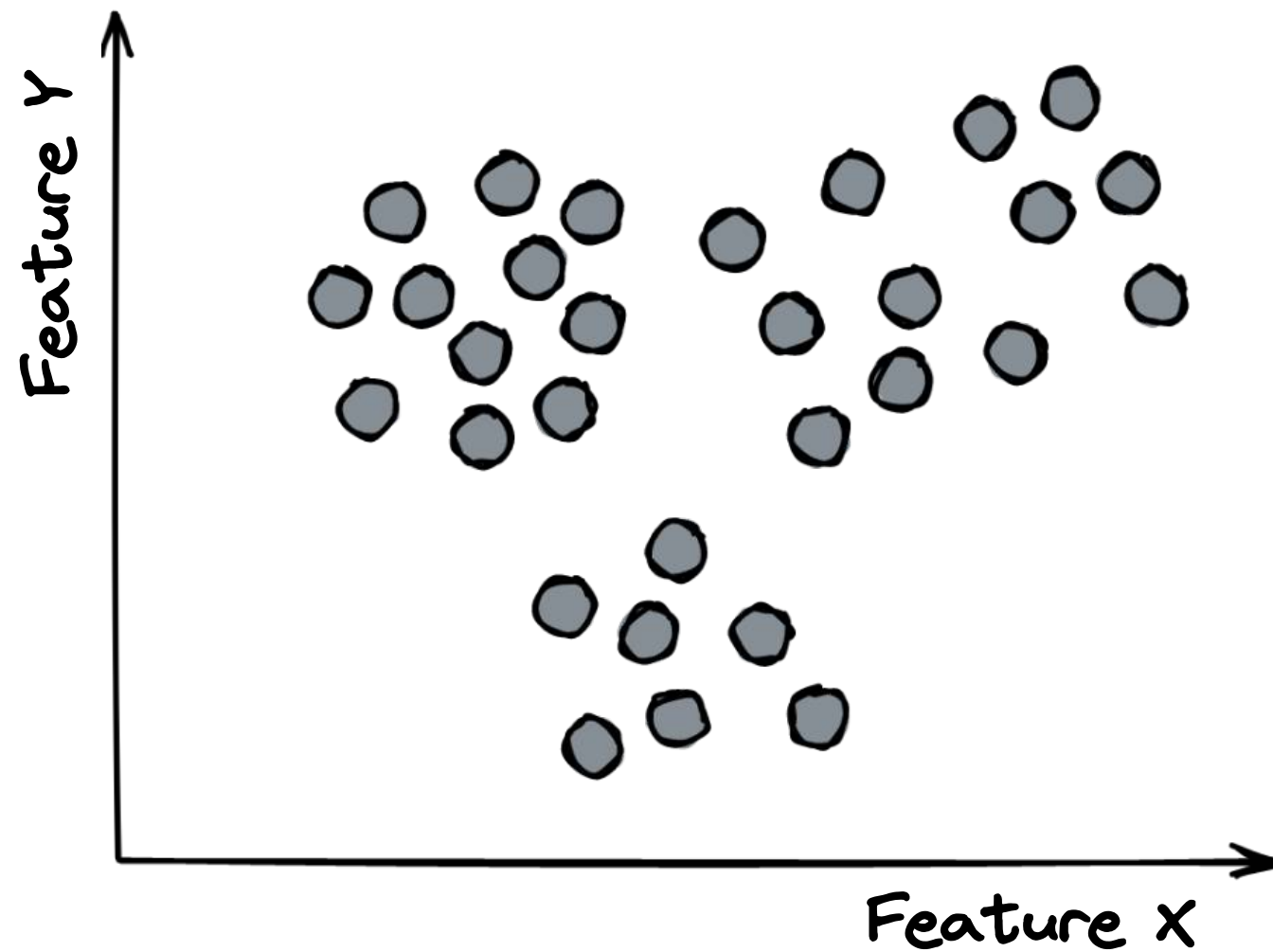
# Clustering



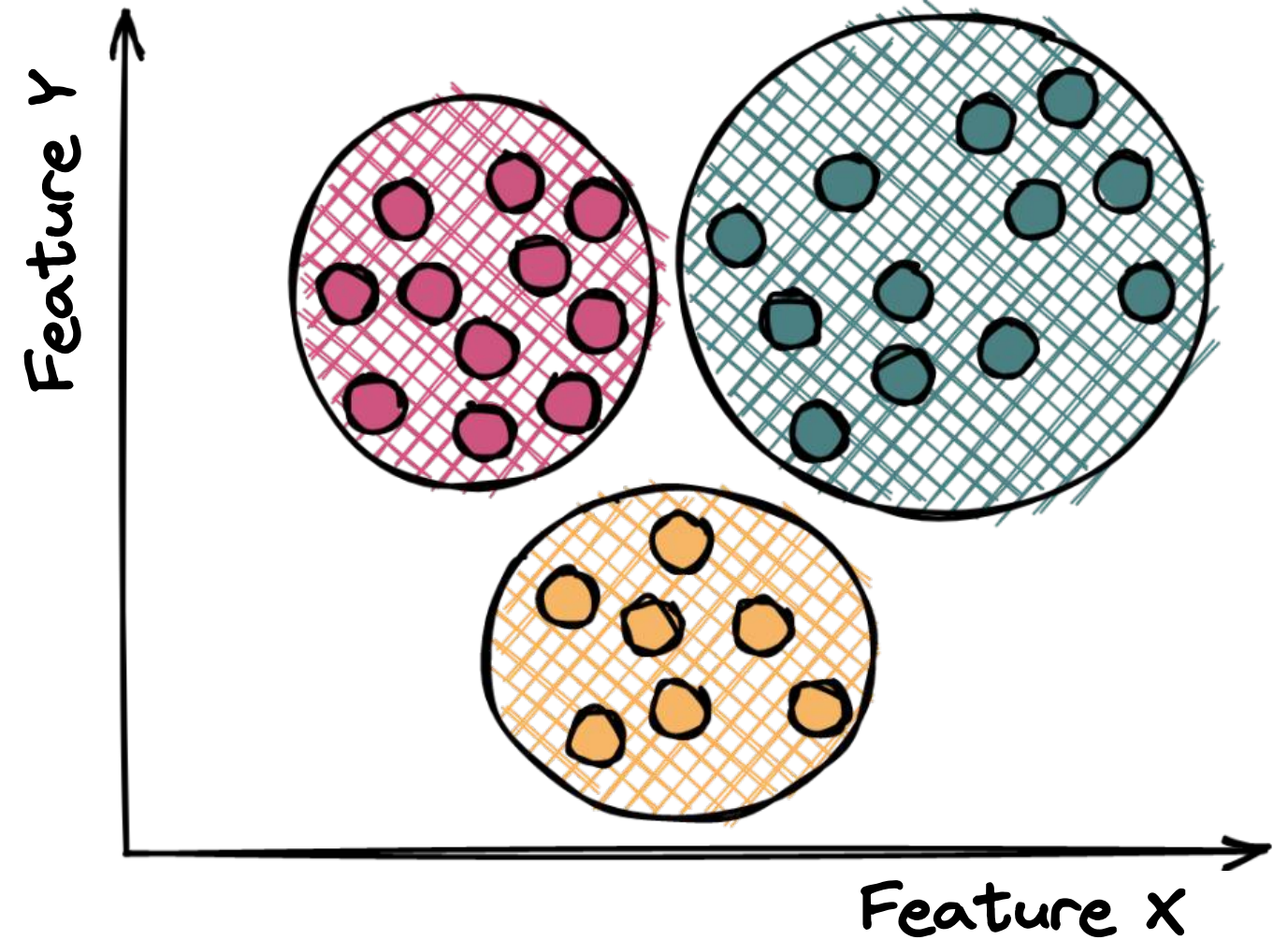
Additional materials

Middlesex University Dubai; CST4050;  
Instructor: Ivan Reznikov

# Unlabeled Data

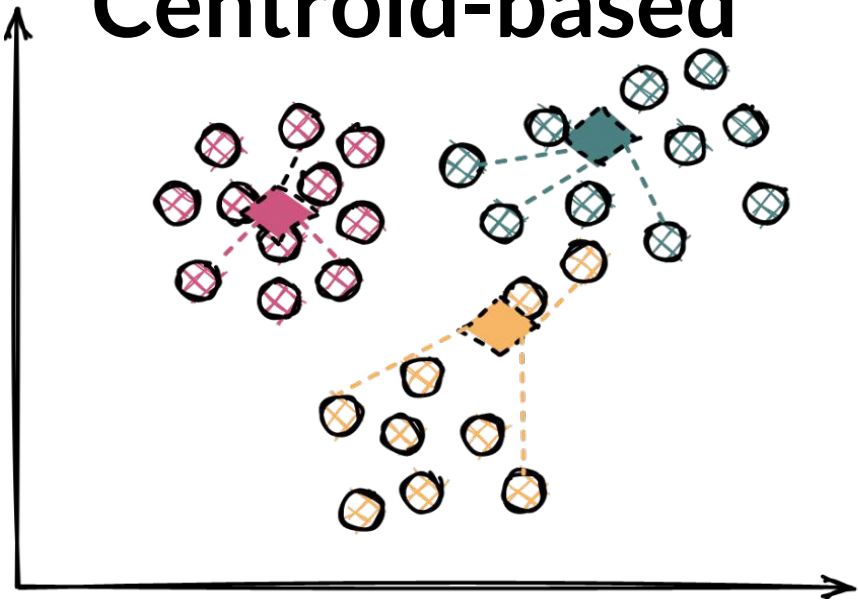


# Labeled Clusters

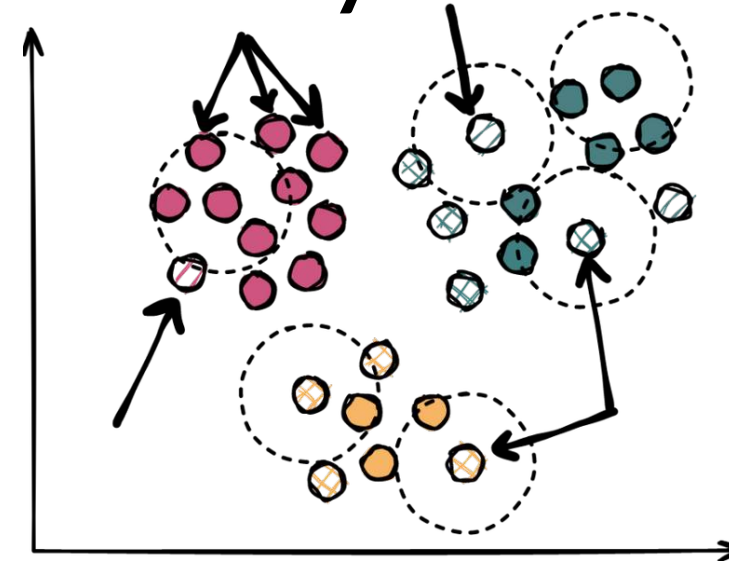


# Types of clustering:

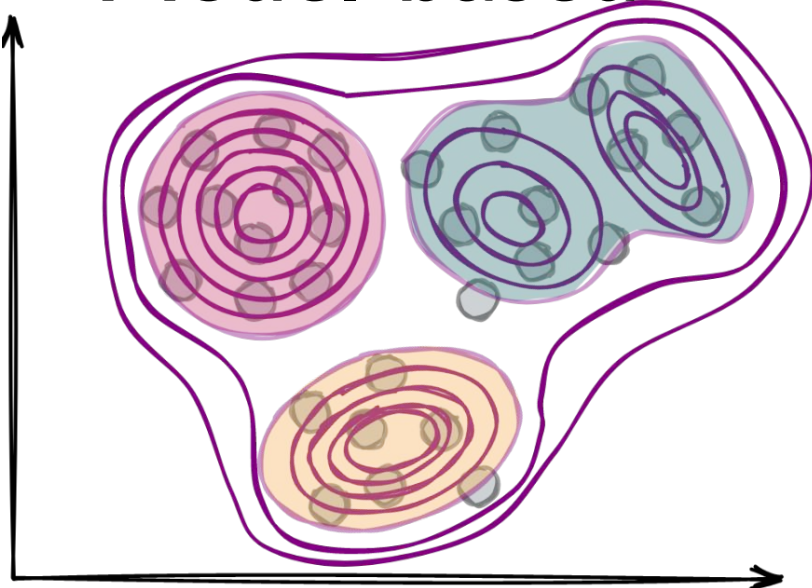
**Centroid-based**



**Density-based**



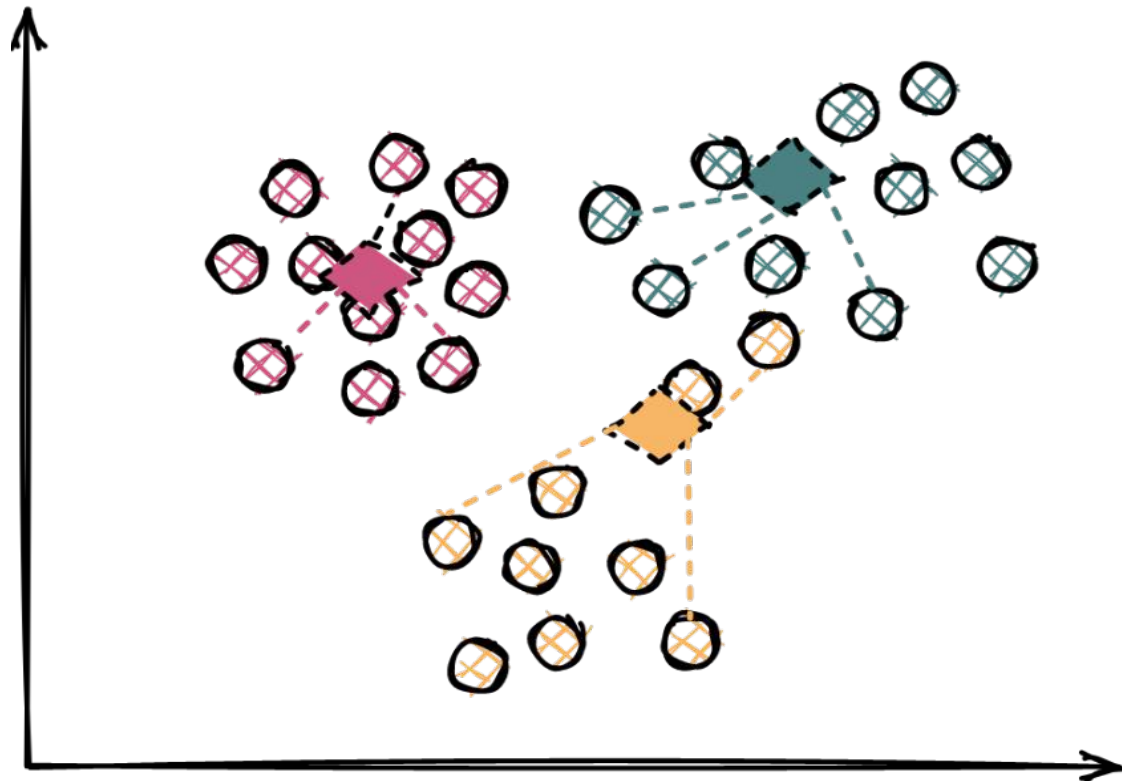
**Model-based**



**Distance-based**



# Centroid-based clustering



**Main idea:**

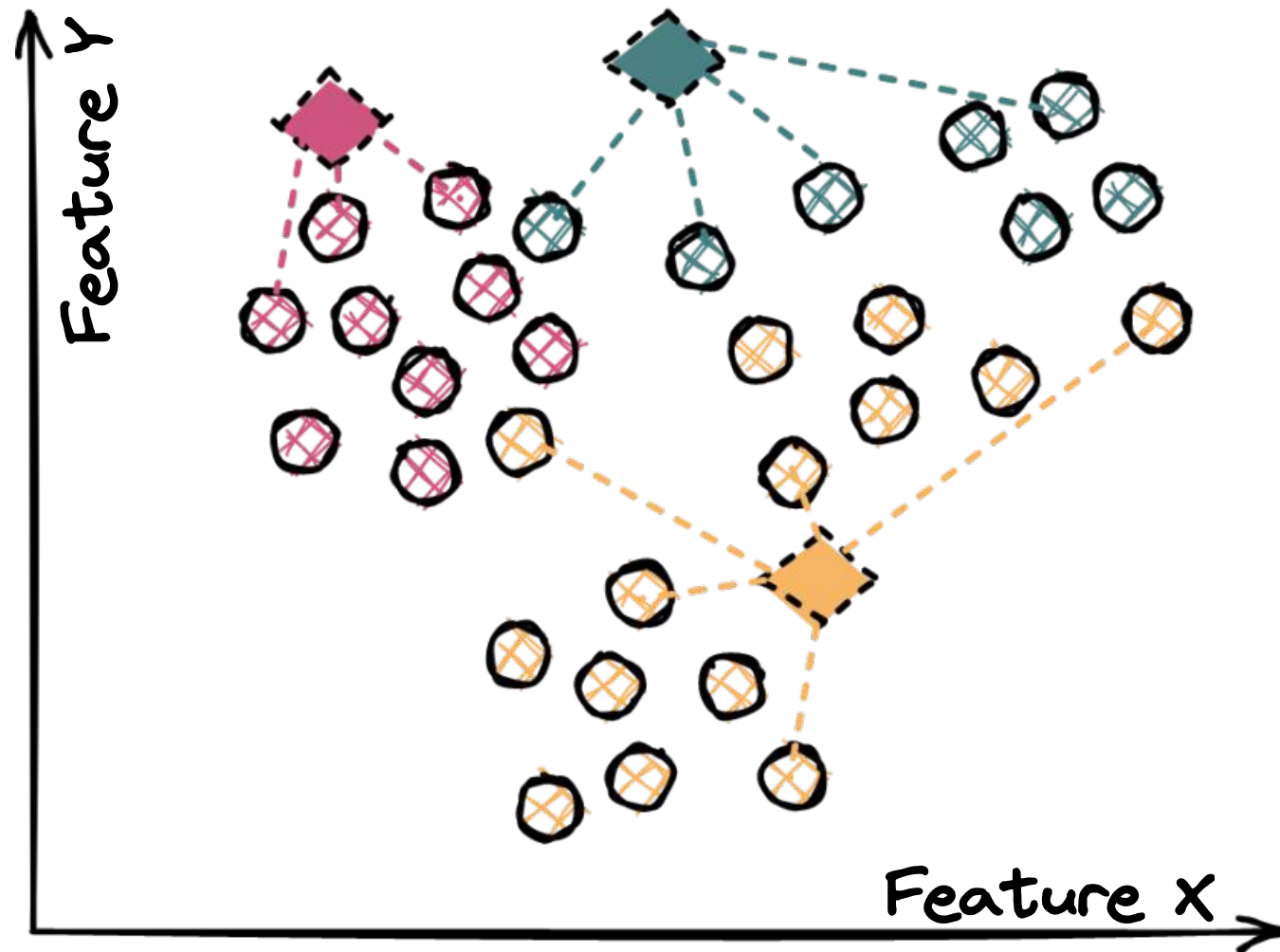
Minimize the squared distances of all points in the cluster to cluster centroids.

**Most used method:**  
k-Means

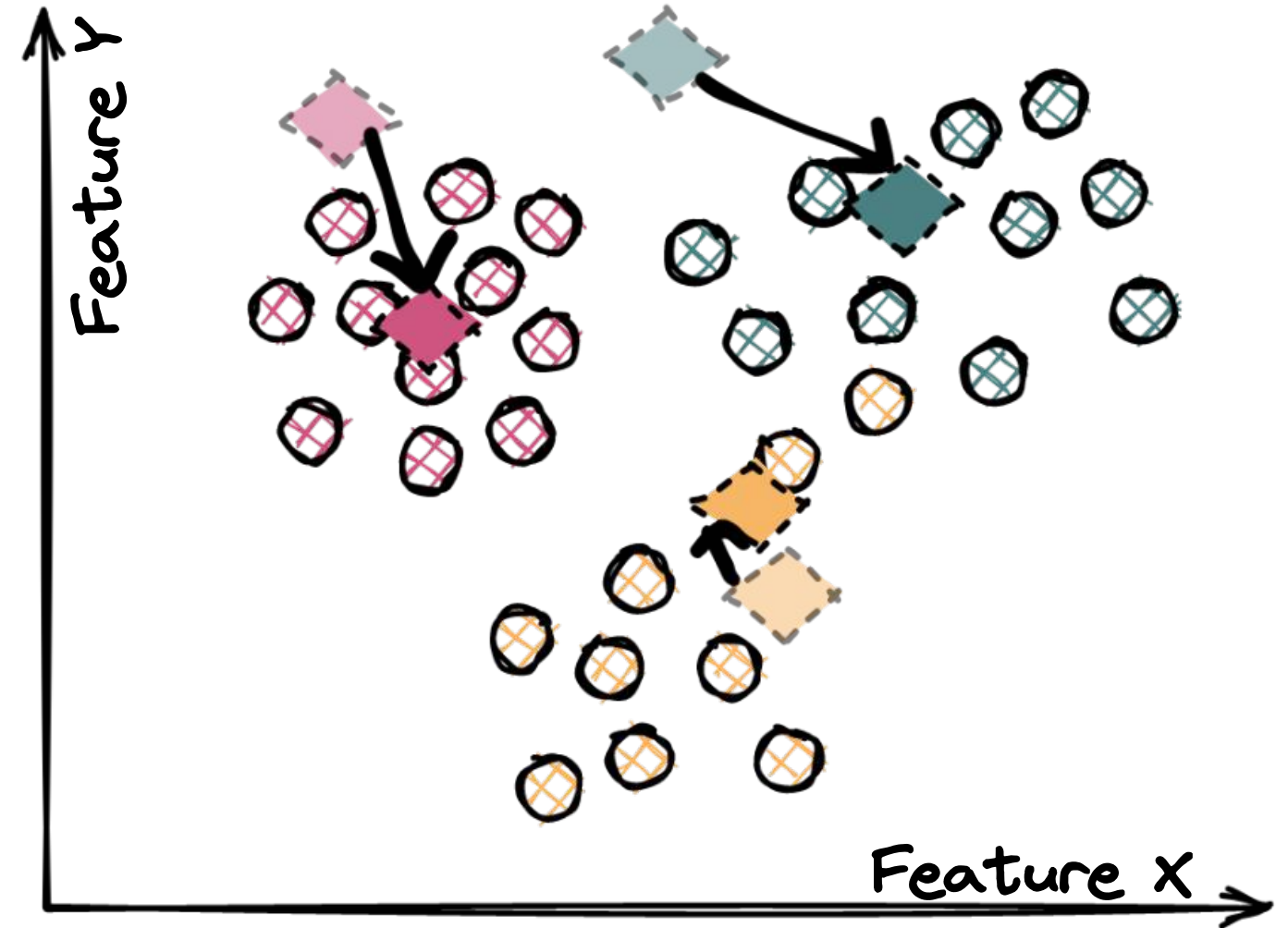


# Centroid-based clustering

Step1a. Calculate distances



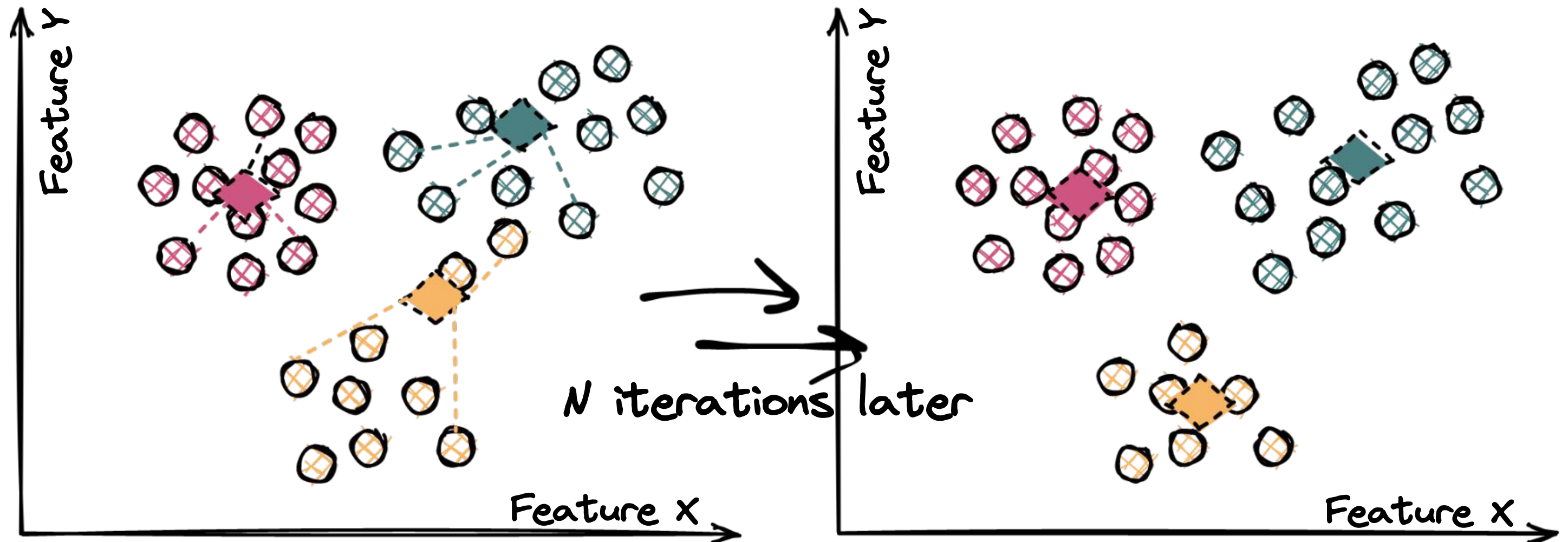
Step1b. Relocate centroids



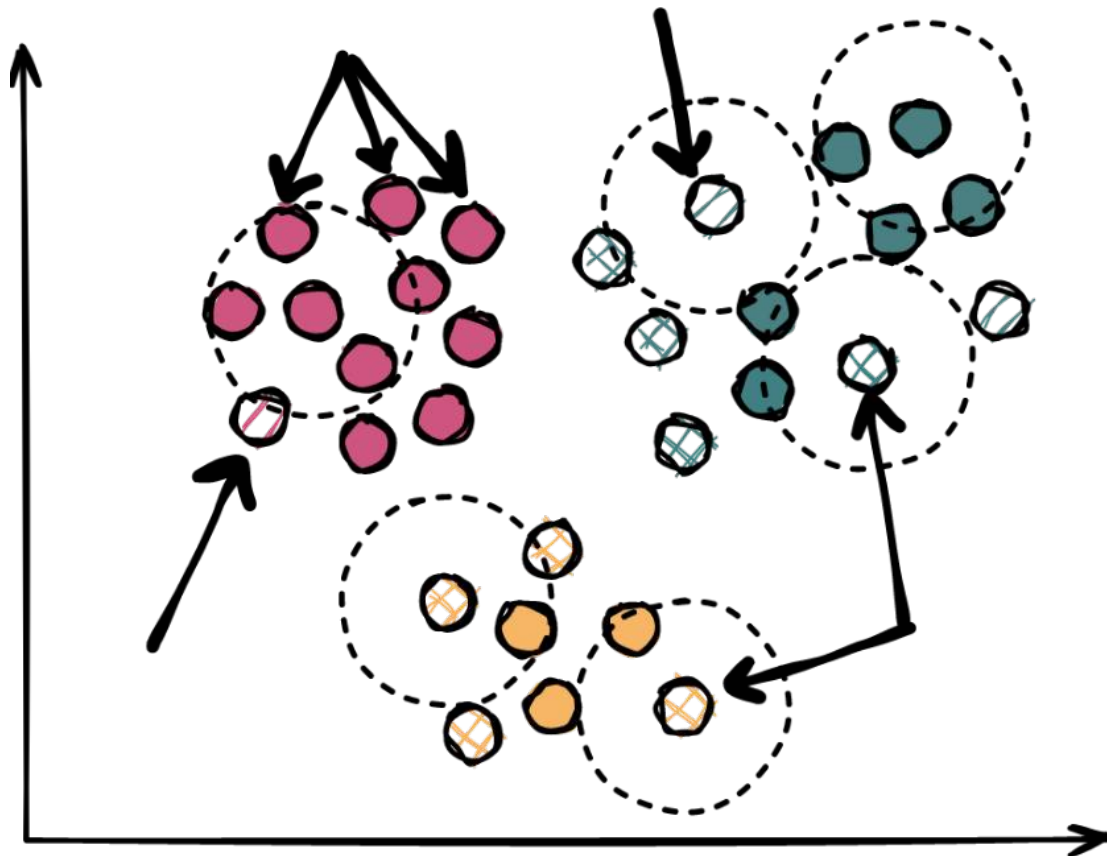
# Centroid-based clustering

Step2a. Calculate distances

StepNb. Relocate centroids



# Density-based clustering



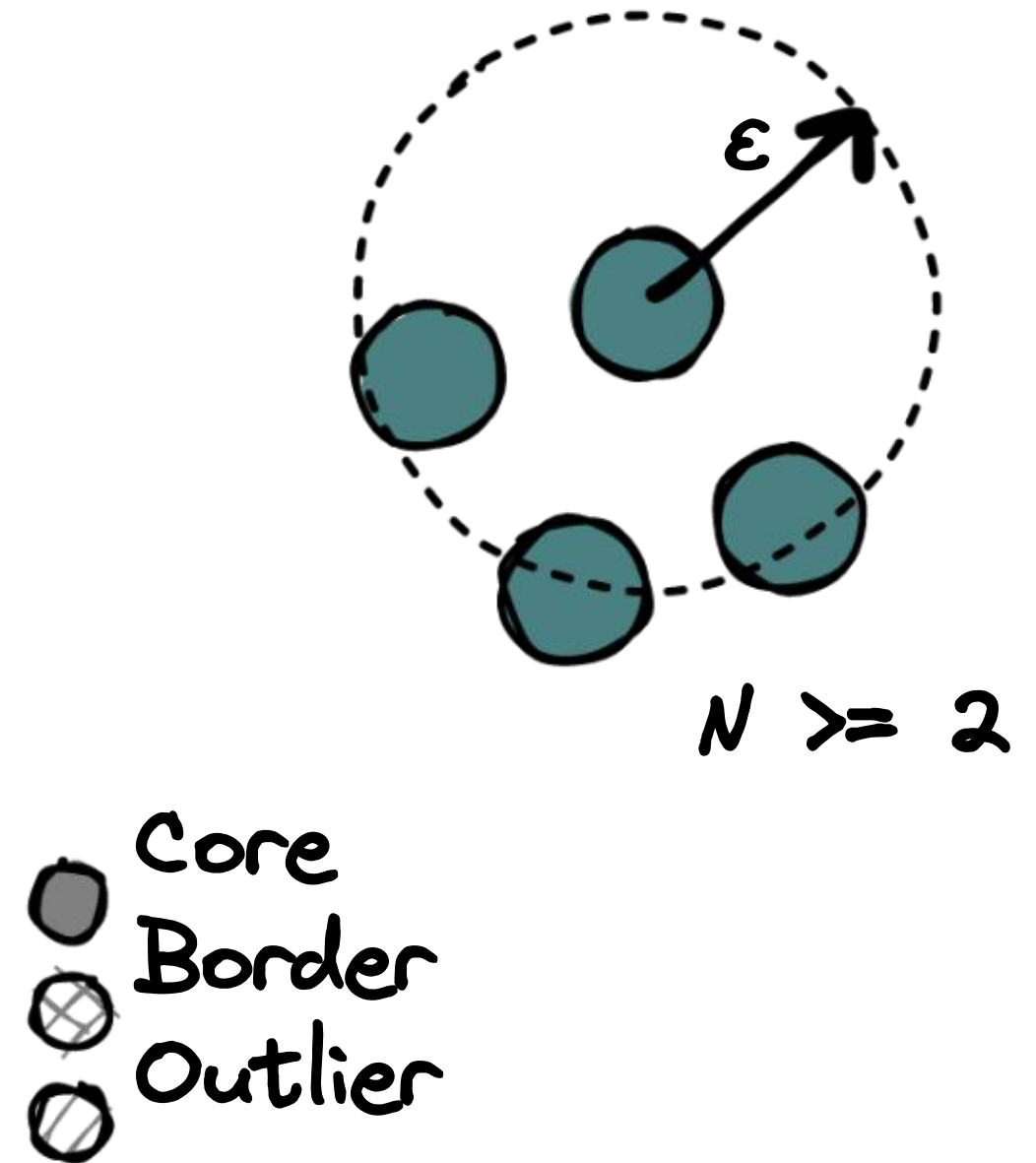
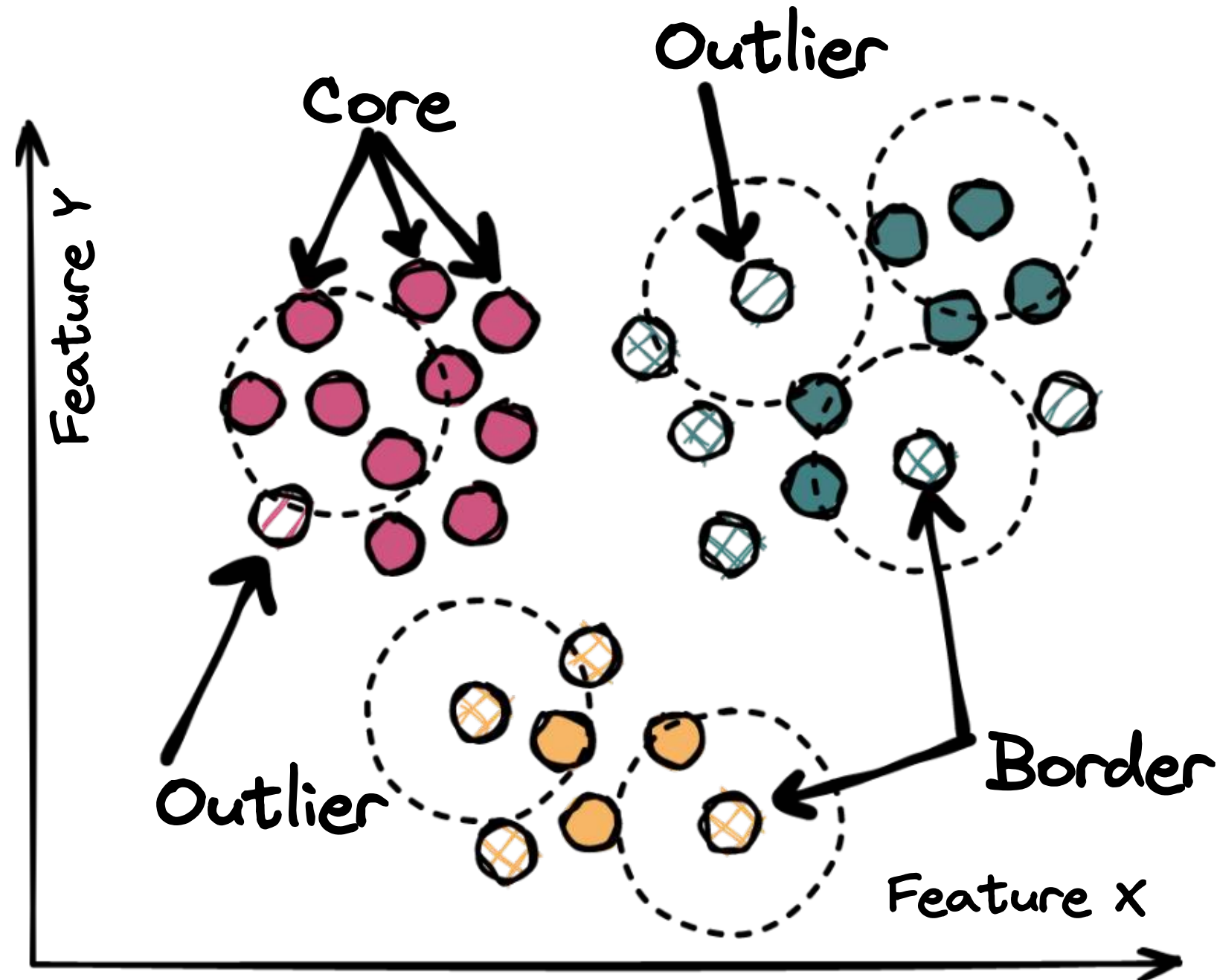
**Main idea:**

Clusters are defined based on identifying areas of higher density.

**Most used method:**

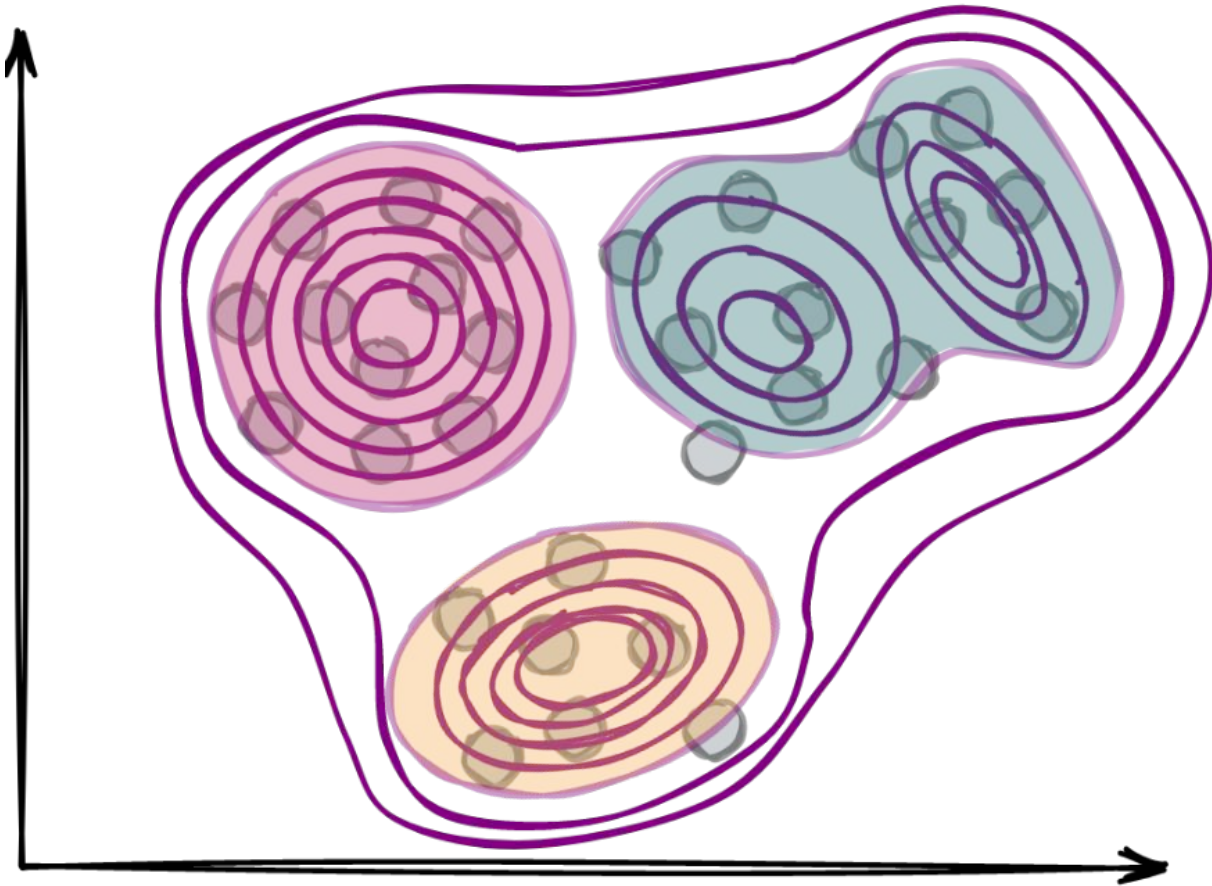
DBSCAN

# Density-based clustering





# Model-based clustering



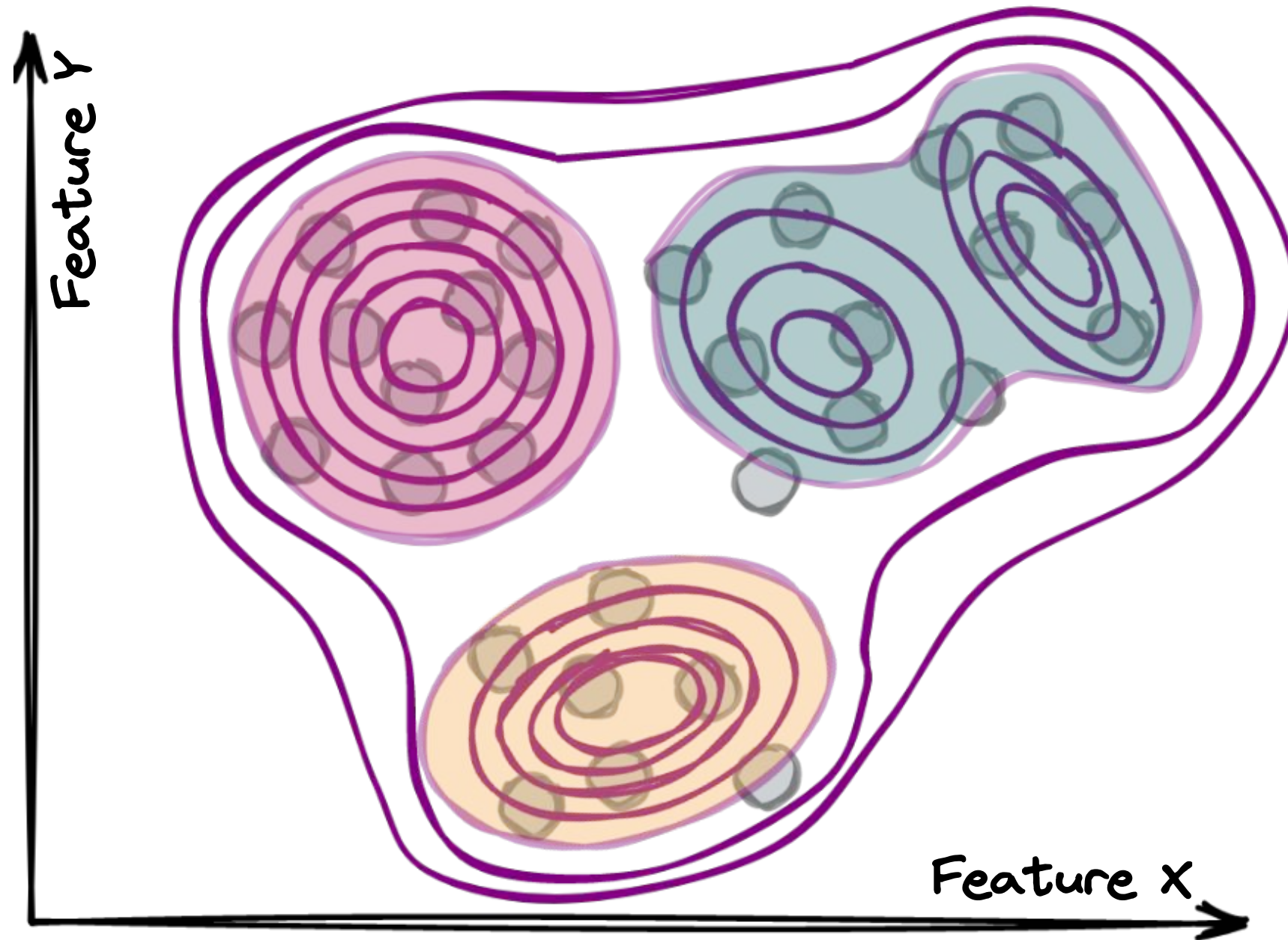
## **Main idea:**

Clusters are defined based on how likely the objects included are likely to belong to the same distribution.

## **Most used method:**

GMM – Gaussian Mixture Models

# Types of clustering: Model-based



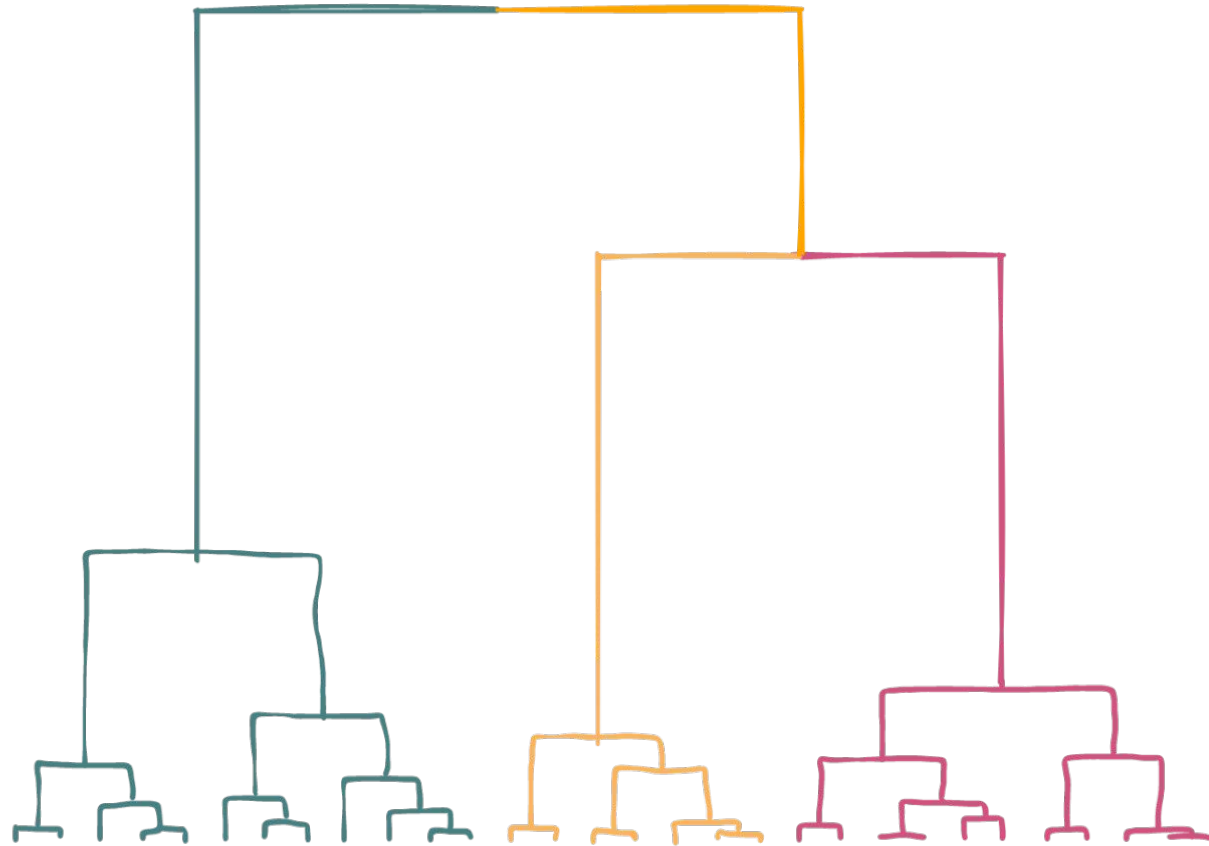
# Distance-based clustering

## Main idea:

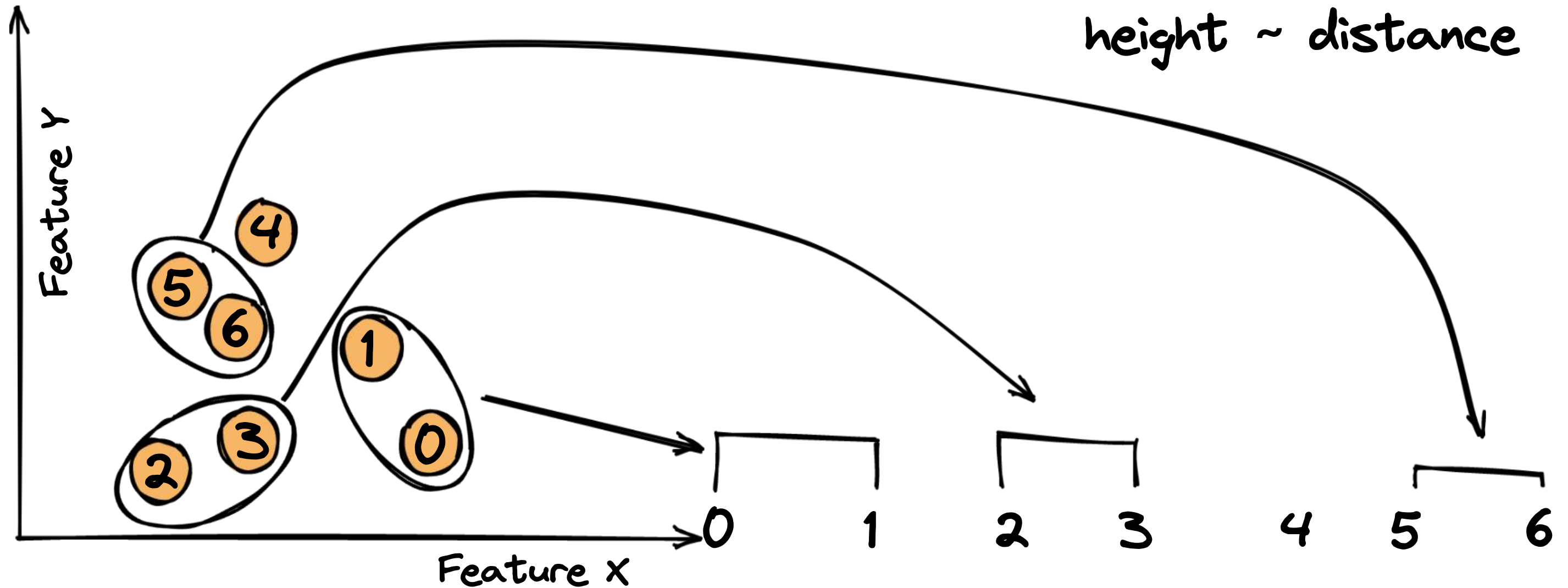
Clusters are developed based on distance between objects, as closer means more related.

## Most used method:

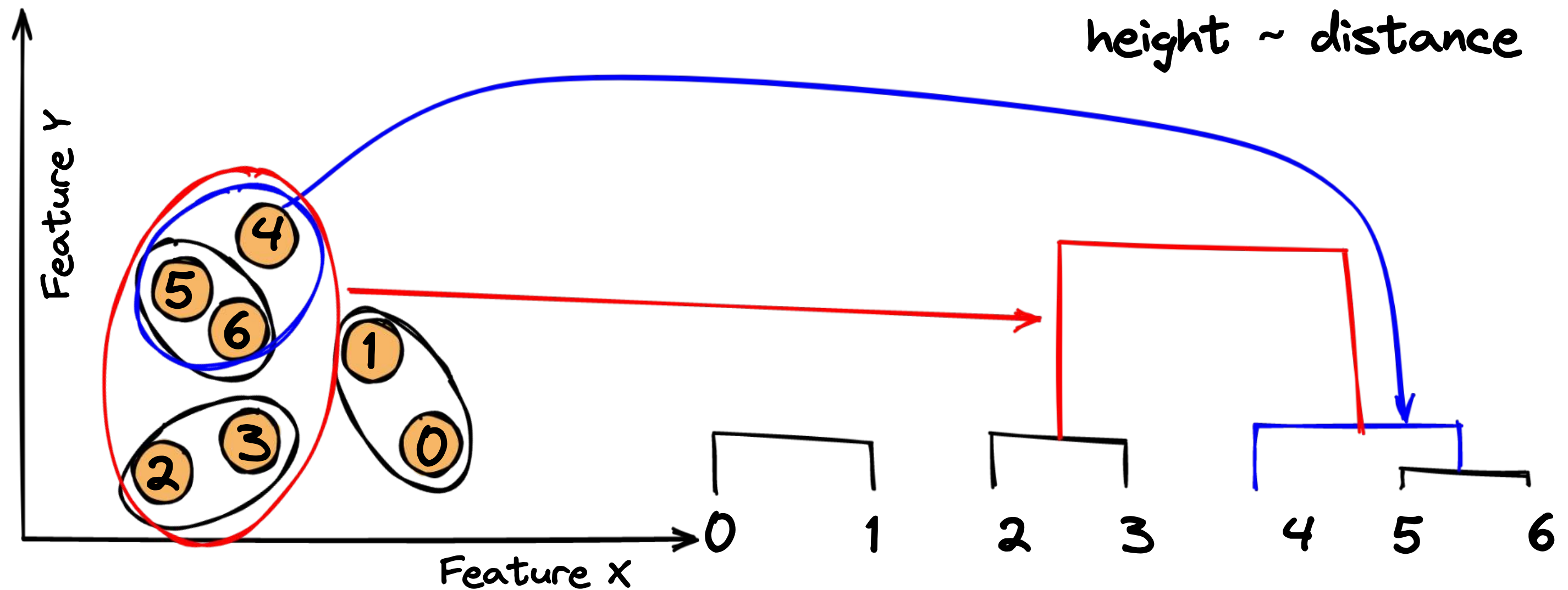
AHC – Agglomerative hierarchical clustering



# Distance-based clustering

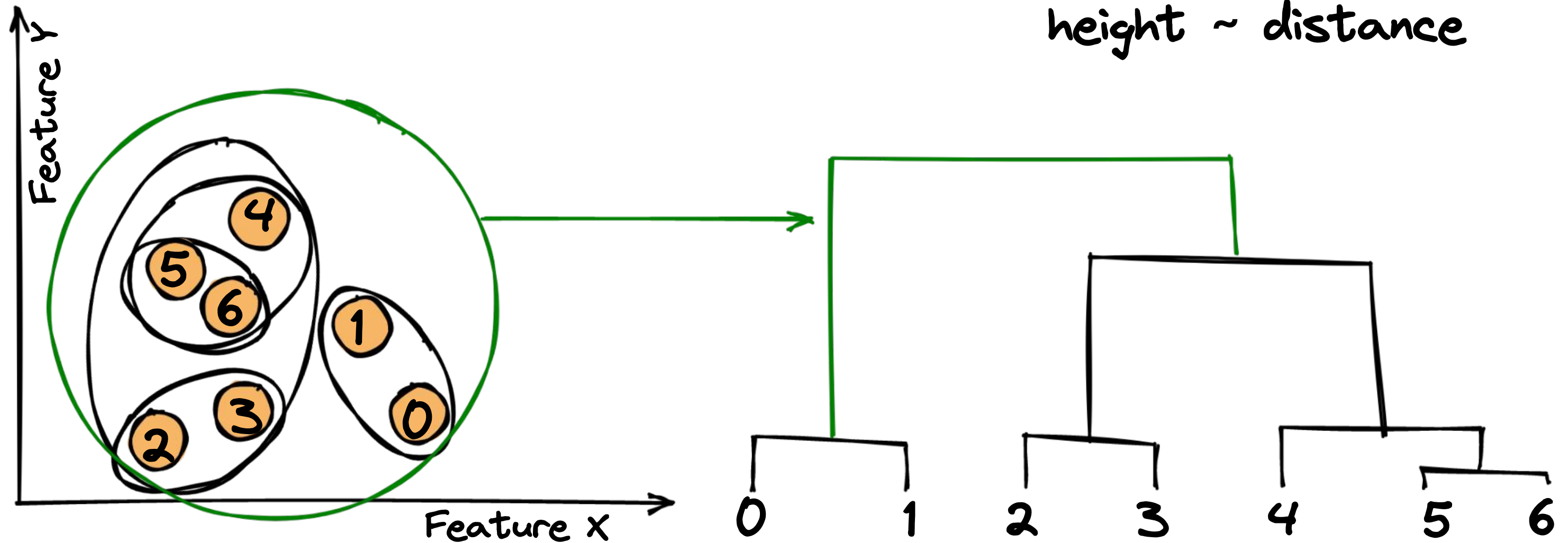


# Distance-based clustering

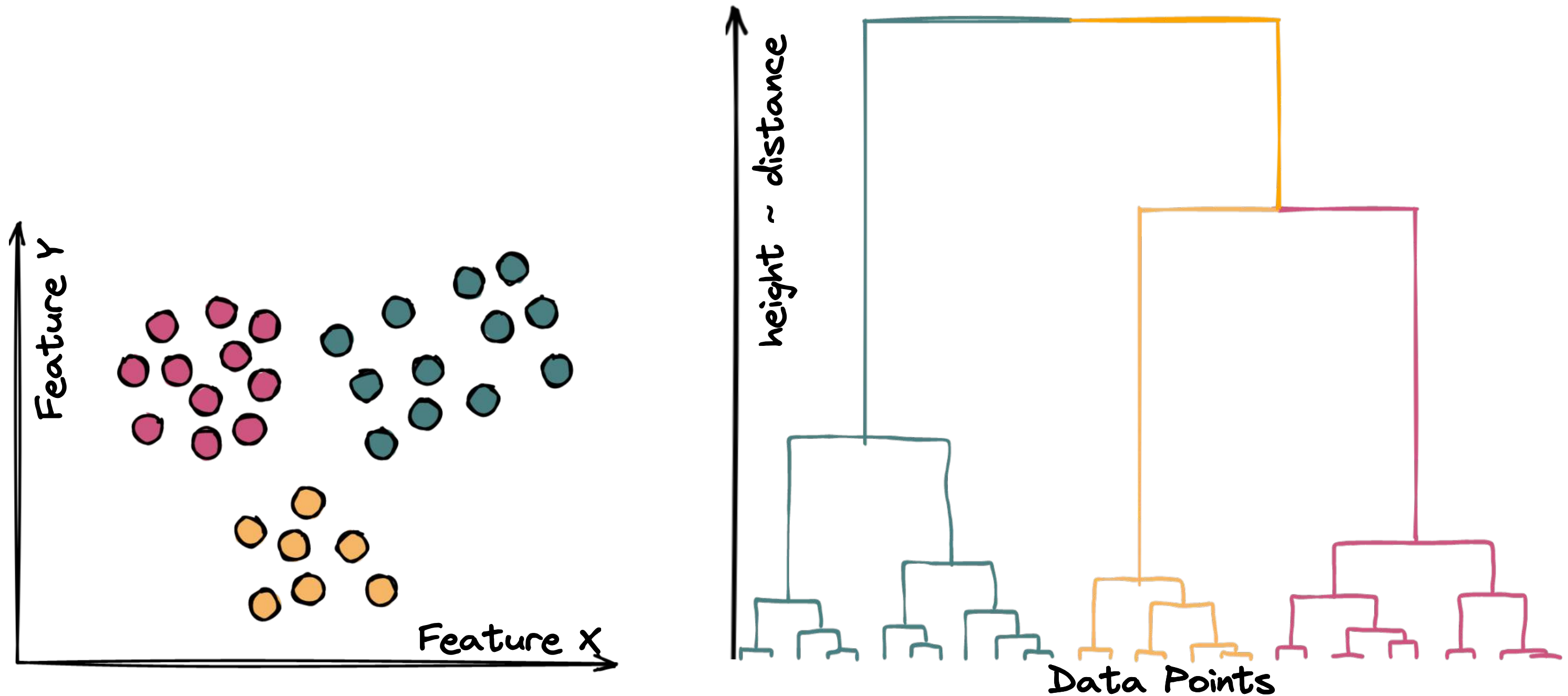











# Distance-based clustering


















# Distance-based clustering



# Advantages

	Centroid	Density	Model	Distance
performance and scaling				
can return probability of points belonging to cluster K				
overlapping clusters can be identified as several				
can work with weird-shaped clusters				
can find clusters surrounded by other clusters				
can provide object ordering				
can return dendrogram				

# Disadvantages

	Centroid	Density	Model	Distance
required K number of cluster				
sensitive to chosen inputs				
scaling problems with high dimensions				
strongly dependent on random				
varying sizes and densities problems				
exposed to noise and outliers				
fails if sparse data				
requires a large amount of data				
needs to know the type of distribution				
can't regroup clusters if done wrong				