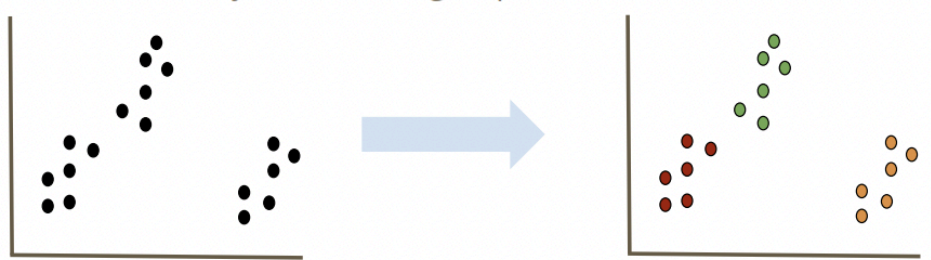


Clustering - Kmeans

1. What is Clustering

- a. A clustering is a grouping / assignment of objects (data points) such that objects in the same group / cluster are
 - i. similar to one another
 - ii. dissimilar to objects in other groups

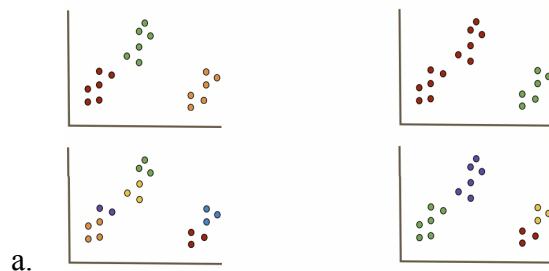
iii.



2. Applications

- a. Outlier detection / anomaly detection
 - i. Data Cleaning / Processing
 - ii. Credit card fraud, spam filter, etc.
- b. Feature Extraction
- c. Filling gaps in the data
 - i. Using the same marketing strategy for similar people
 - ii. Infer probable values for gaps in the data (similar users could have similar hobbies, likes / dislikes, etc.)

3. Clusters can be Ambiguous

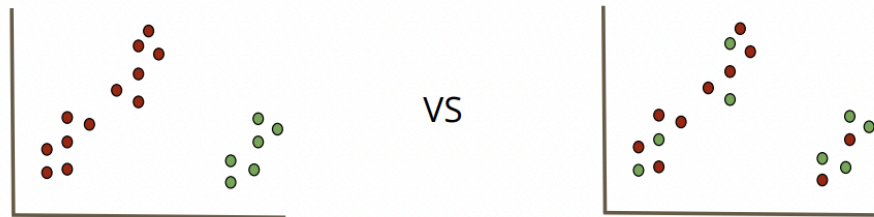


4. Types of Clusterings

- a. Partitional
 - i. Each object belongs to exactly one cluster
- b. Hierarchical
 - i. A set of nested clusters organized in a tree
- c. Density-Based
 - i. Defined based on the local density of points
- d. Soft Clustering
 - i. Each point is assigned to every cluster with a certain probability

5. Partitional Clustering

- Goal: partition dataset into k partitions



b.

6. Example

- Given a distance function d , we can find points (not necessarily part of our dataset) for each cluster called centroids that are at the center of each cluster.



b.

- Q: When d is Euclidean, what is the centroid (also called center of mass) of m points $\{x_1, \dots, x_m\}$?
- Looking at the sum of the distances of points in a cluster to its centroid also captures the “spread” (variance) of a cluster

ance) of a cluster

$$\sum_i^k \sum_{x \in C_i} d(x, \mu_i)^2$$

Mean of cluster i

Cluster i

7. Cost Function

- Way to evaluate and compare solutions
- Hope: can find some algorithm that find solutions that make the cost small
- Q: Can you suggest a cost function to use for partitional clustering?

$$\sum_i^k \sum_{x \in C_i} d(x, \mu_i)^2$$

8. K-means

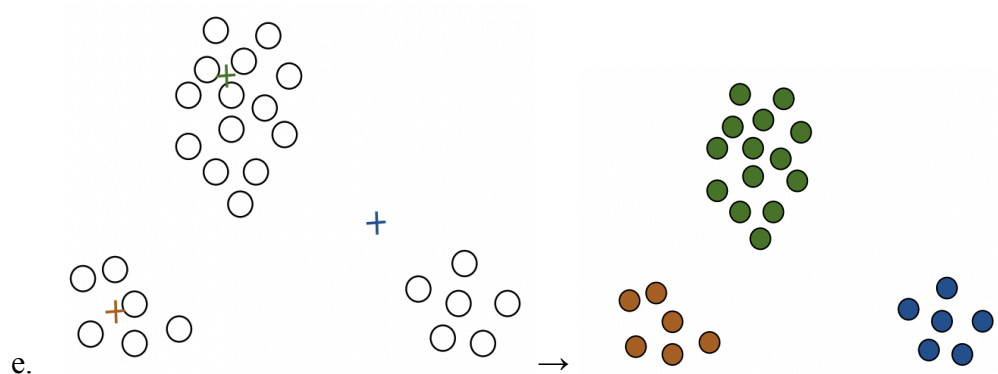
- Given $X = \{x_1, \dots, x_n\}$ our dataset and k
- Find k points $\{u_1, \dots, u_k\}$ that minimize the cost function:

$$\sum_i^k \sum_{x \in C_i} d(x, \mu_i)^2$$

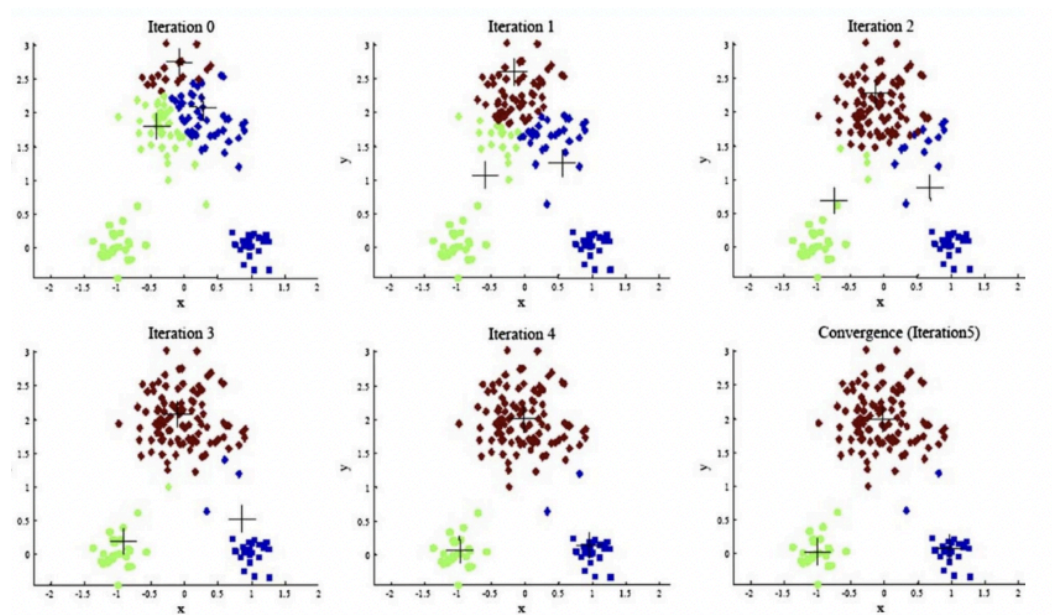
- c. When $k = 1$ and $k = n$ this is easy. Why?
 - i. If $k = n$, every data point is a cluster
 - ii. If $k = 1$, the whole data point is a cluster
- d. When x_i lives in more than 2 dimensions, this is a very difficult (NP-hard) problem

9. K-means - Lloyd's Algorithm

- a. Randomly pick k centers $\{u_1, \dots, u_k\}$
- b. Assign each point in the dataset to its closest center
- c. Compute the new centers as the means of each cluster
- d. Repeat 2 & 3 until convergence



e.



f.