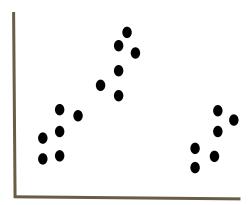
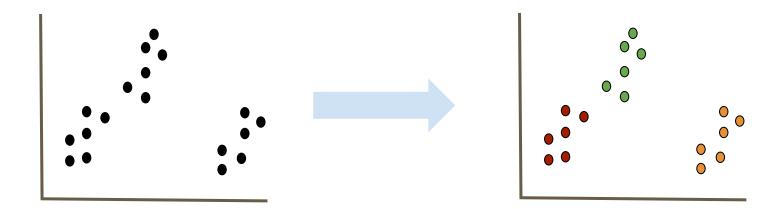
# **Clustering - Kmeans**

Boston University CS 506 - Lance Galletti

# What is a Clustering



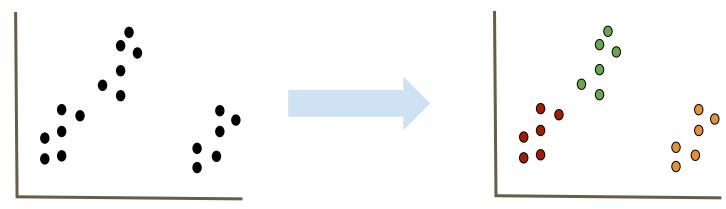
# What is a Clustering



### What is a Clustering

A clustering is a grouping / assignment of objects (data points) such that objects in the same group / cluster are:

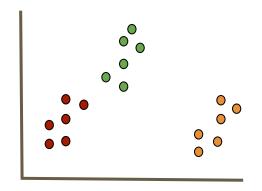
- similar to one another
- dissimilar to objects in other groups

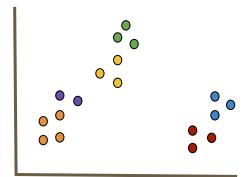


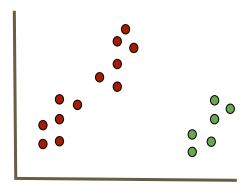
### **Applications**

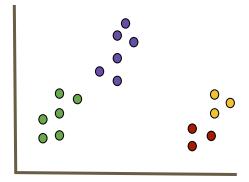
- Outlier detection / anomaly detection
  - Data Cleaning / Processing
  - Credit card fraud, spam filter etc.
- Feature Extraction
- Filling Gaps in your data
  - Using the same marketing strategy for similar people
  - Infer probable values for gaps in the data (similar users could have similar hobbies,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

# **Clusters can be Ambiguous**









### **Types of Clusterings**

#### **Partitional**

Each object belongs to exactly one cluster

#### Hierarchical

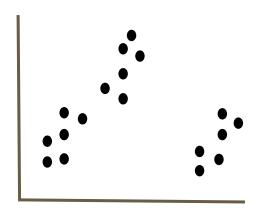
A set of nested clusters organized in a tree

#### **Density-Based**

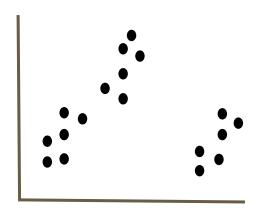
Defined based on the local density of points

#### **Soft Clustering**

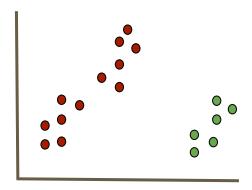
Each point is assigned to every cluster with a certain probability

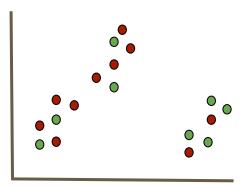


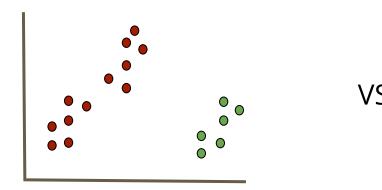


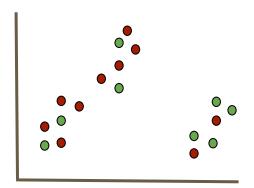












#### **Example**



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.

#### **Example**



Q: When **d** is Euclidean, what is the **centroid** (also called **center of mass**) of **m** points  $\{x_1, ..., x_m\}$ ?

A: The mean / average of the points

#### **Example**



Looking at the sum of the distances of points in a cluster to its centroid also captures the "spread" (variance) of a cluster

$$\sum_{i}^{k} \sum_{x \in C_{i}} \operatorname{d}(\mathbf{x}, \mu_{\mathbf{i}})^{2}$$
 Cluster i

#### **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?

$$\sum_{i}^{\kappa} \sum_{x \in C_{i}} d(x, \mu_{i})^{2}$$

#### K-means

Given  $X = \{x_1, ..., x_n\}$  our dataset and k

Find **k** points  $\{\mu_1, ..., \mu_k\}$  that minimize the **cost function**:

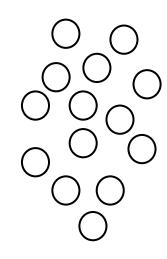
$$\sum_{i}^{k} \sum_{x \in C_{i}} d(x, \mu_{i})^{2}$$

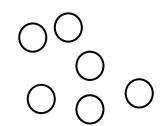
When **k=1** and **k=n** this is easy. Why?

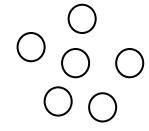
When  $\mathbf{x_i}$  lives in more than 2 dimensions, this is a very difficult (**NP-hard**) problem

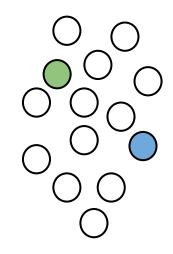
#### K-means - Lloyd's Algorithm

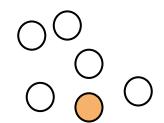
- 1. Randomly pick **k** centers  $\{\mu_1, ..., \mu_k\}$
- 2. Assign each point in the dataset to its closest center
- 3. Compute the new centers as the means of each cluster
- 4. Repeat 2 & 3 until convergence

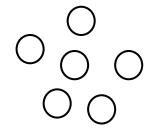


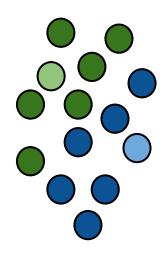


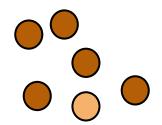


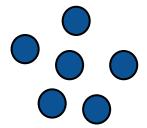


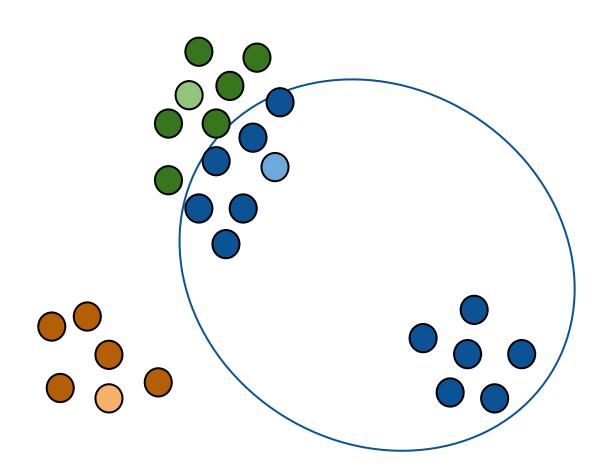


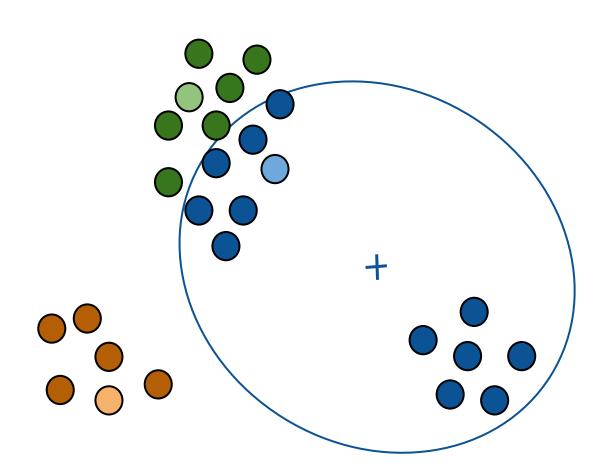


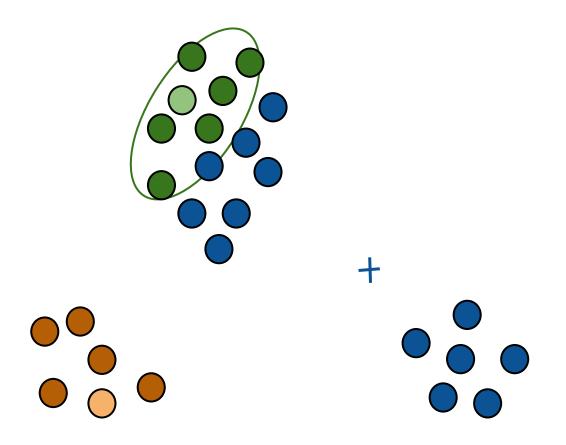


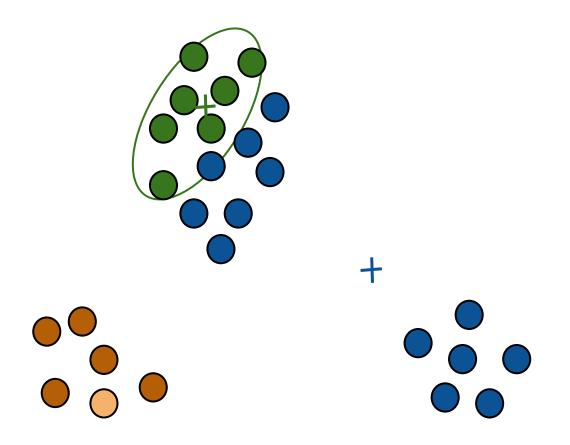


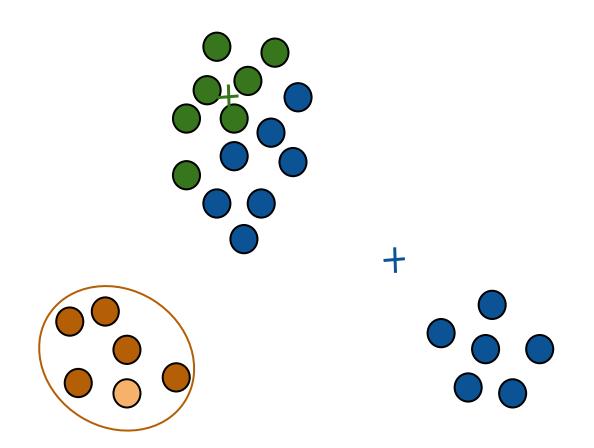


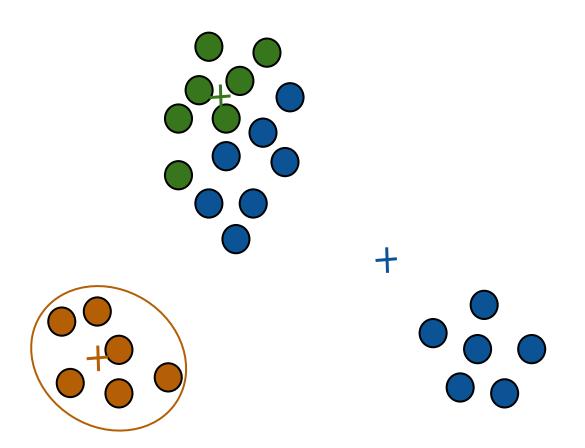


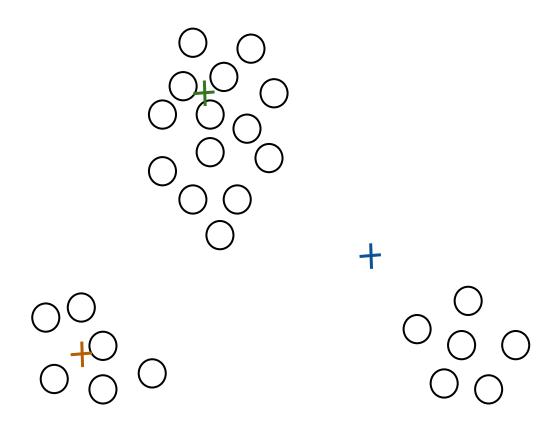


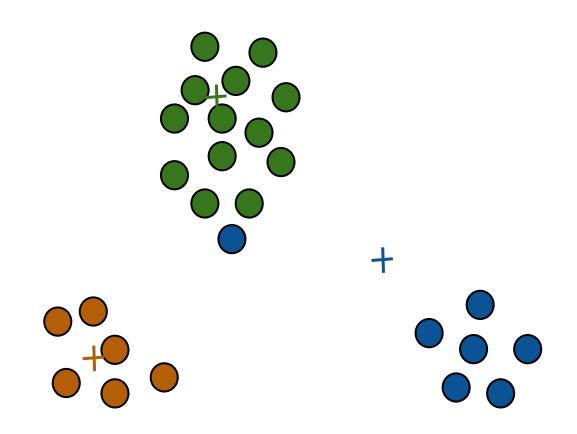


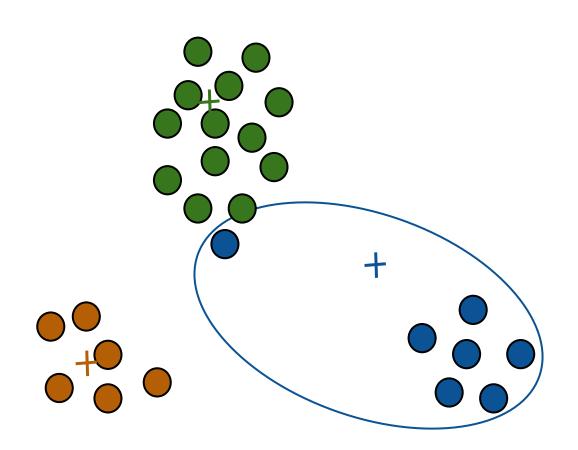


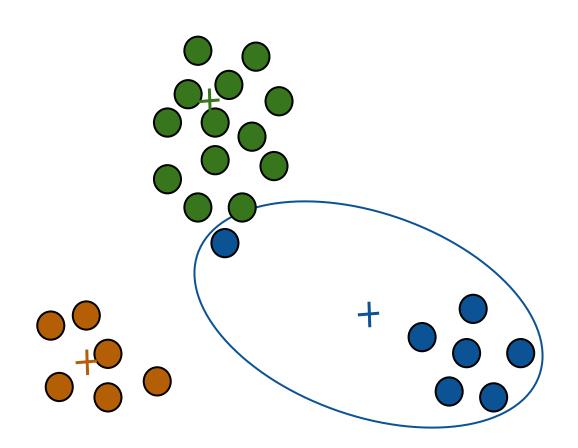


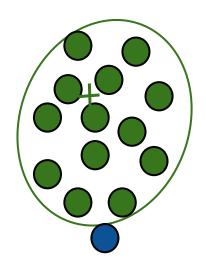


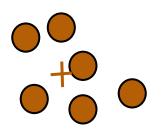


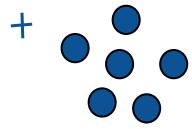


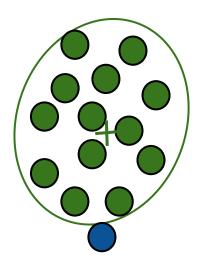


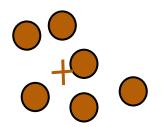


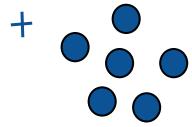


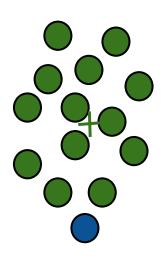


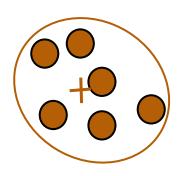


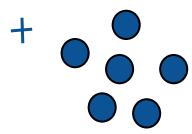


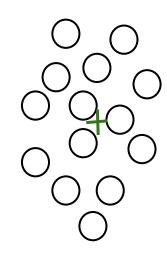


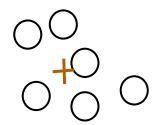


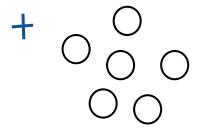


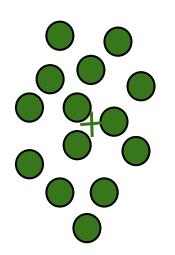


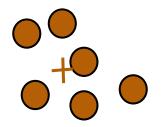


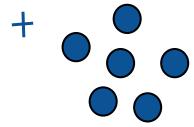


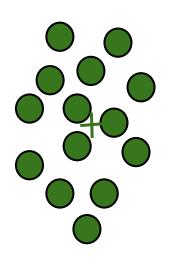


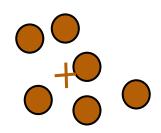


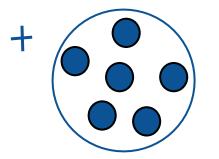


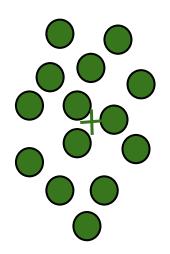


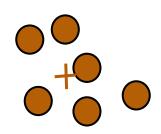


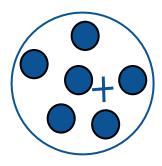


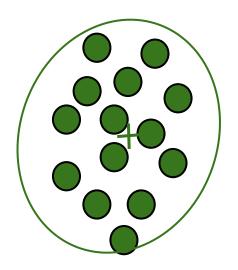


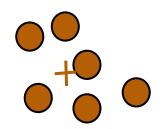


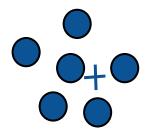


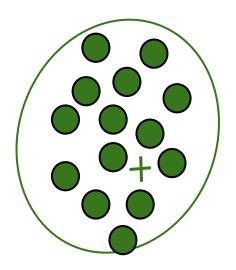


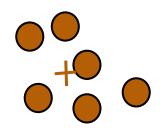


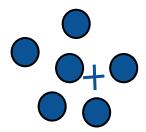


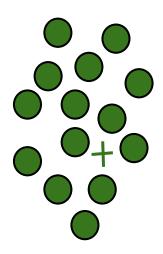


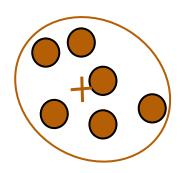


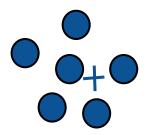


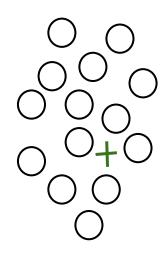


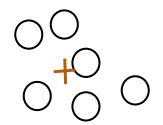


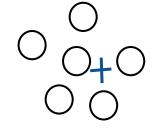


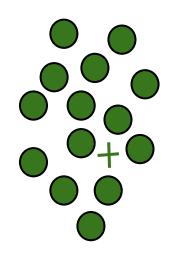


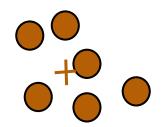


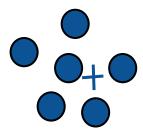


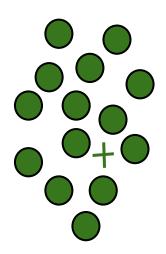


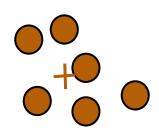


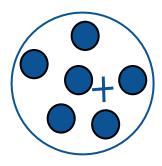


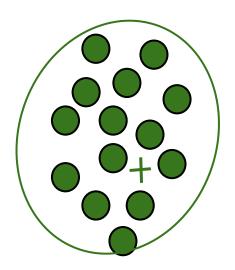


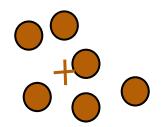


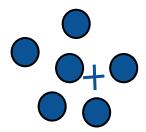


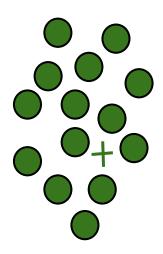


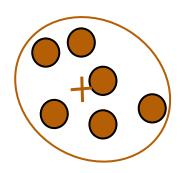


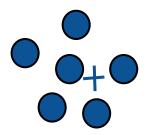


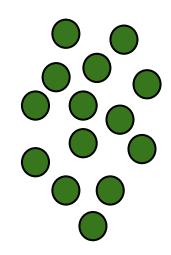


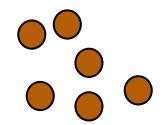


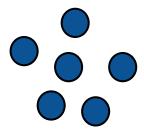




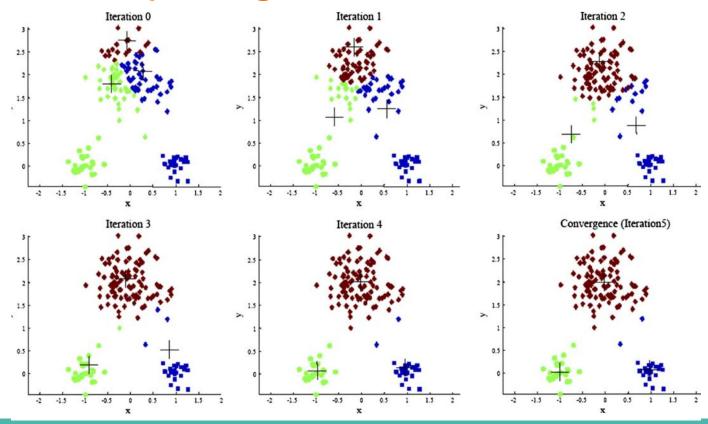








## K-means - Lloyd's Algorithm



## **Worksheet-5min**

Please do a) -> d) of the worksheet with the person sitting next to you.

## **Worksheet - 5min**

Share your answers with the group next to you. Discuss / debate if you have different answers.