

Lecture 9: Clustering

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AGENDA

| 01 | Clustering: Overview |
|----|----------------------------------|
| 02 | K-Means Clustering |
| 03 | Hierarchical Clustering |
| 04 | Density-based Clustering: DBSCAN |
| 04 | R Exercise |

- K-Means Clustering (KMC)
 - √ Partitional clustering approach
 - Each cluster is associated with a centroid
 - Each point is assigned to the cluster with the closest centroid
 - Number of cluster, K, must be specified

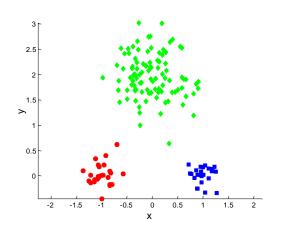
$$\mathbf{X} = C_1 \cup C_2 \dots \cup C_K, \quad C_i \cap C_j = \phi, \quad i \neq j$$

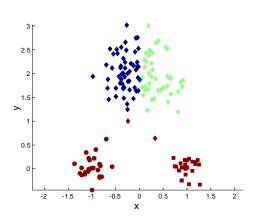
$$\arg \min_{\mathbf{C}} \sum_{i=1}^K \sum_{\mathbf{x} \in C} ||\mathbf{x}_j - \mathbf{c}_i||^2$$

K-Means Clustering Procedure

- 1: Select K points as the initial centroids.
- 2: repeat
- 3: Form K clusters by assigning all points to the closest centroid.
- 4: Recompute the centroid of each cluster.
- 5: **until** The centroids don't change

✓ Initial centroids are often chosen randomly: clustering results vary according to the initial centroid selection



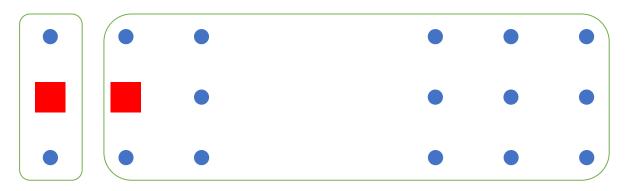


Example

✓ Step 1: Initializing K centroids

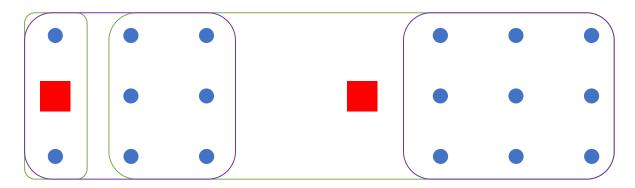


- ✓ Step 2-I (Ist): Assign each instance to the closest center
- ✓ Step 2-2 (Ist): Re-compute the centroids based on the assigned instances

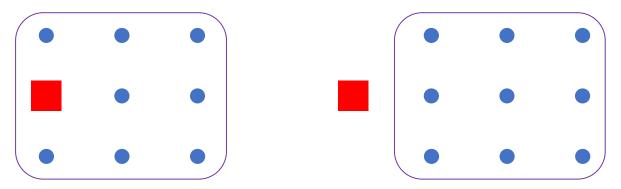


• Example

✓ Step 2-I (2nd): Assign each instance to the closest center



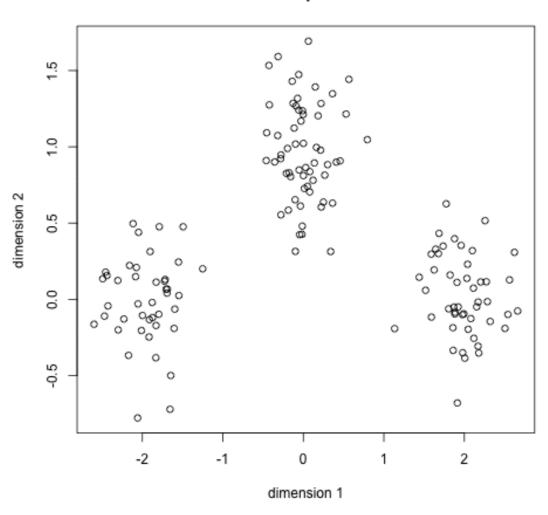
✓ Step 2-2 (2^{nd}): Re-compute the centroids based on the assigned instances



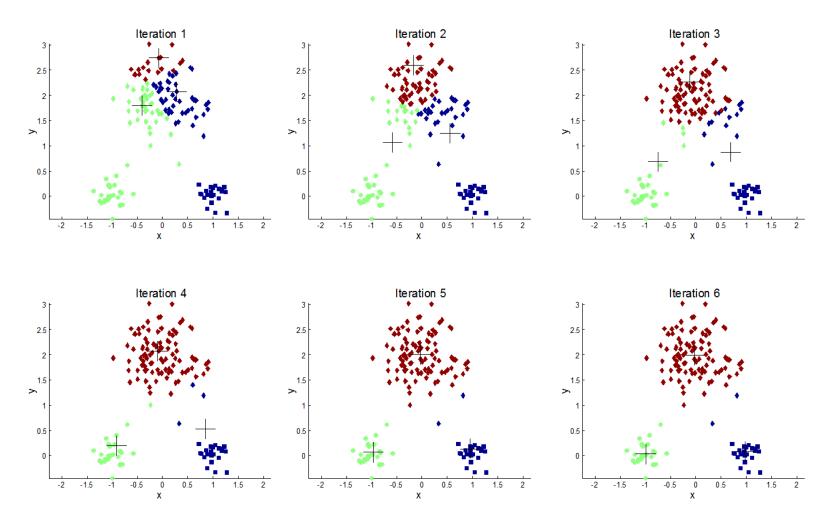
✓ Stop the algorithm because there is no change for centroids and membership
assignment

• KMC example

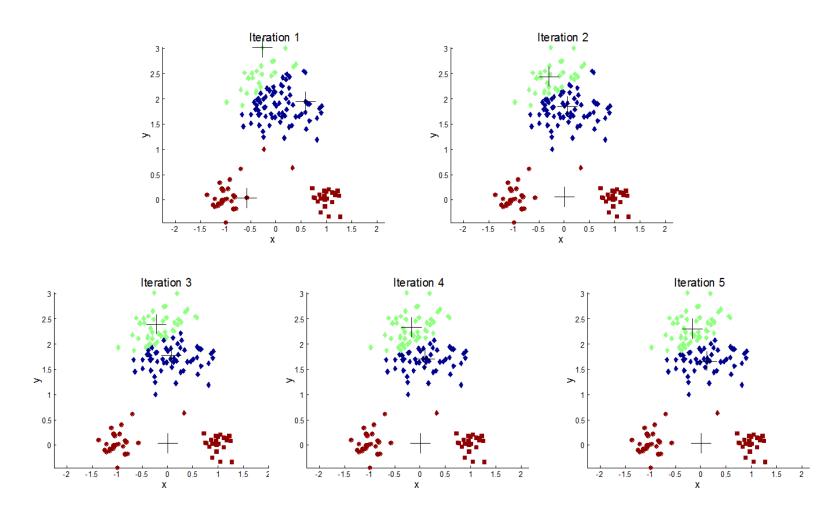




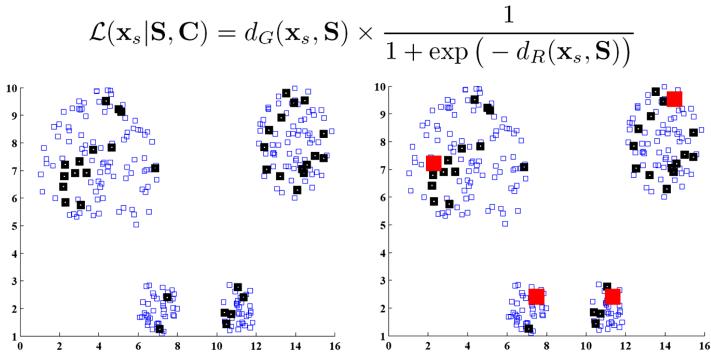
- Effect of initial centroids
 - ✓ Desirable centroid selection



- Effects of initial centroids
 - ✓ Undesirable centroid selection

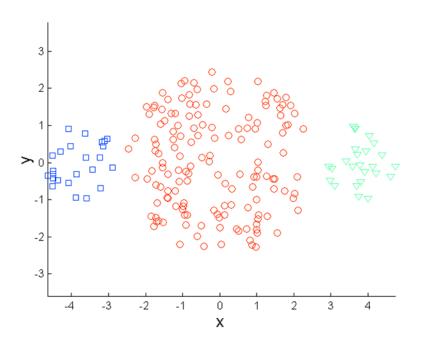


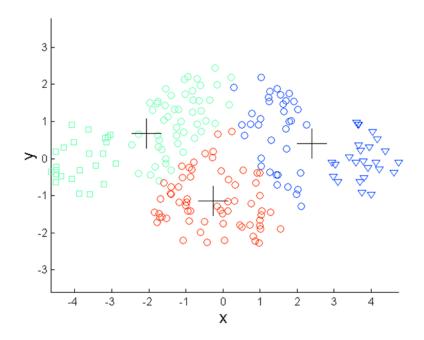
- Some remedies for initial centroid selection
 - ✓ Multiple runs
 - ✓ Sample and use hierarchical clustering to determine initial centroids
 - √ Preprocessing & Postprocessing



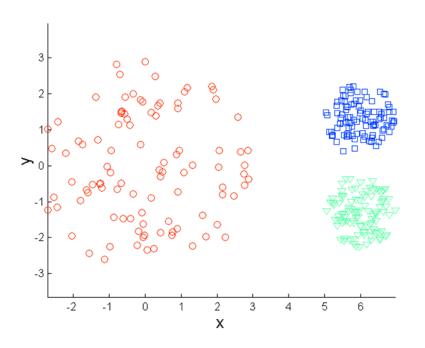
Pilsung Kang and Sungzoon Cho. (2009). K-Means clustering seeds initialization based on centrality, sparsity, and isotropy. *The 13th International Conference on Intelligent Data Engineering and Automated Learning (IDEAL 2009)*, Burgos, Spain. E. Corchado and H. Yin (Eds.), *Lecture Notes in Computer Science LNCS 5788*, 109-117.

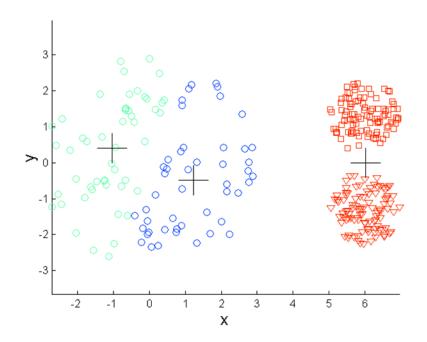
- Limitations of K-Means Clustering
 - ✓ Cannot cope with different sizes





- Limitations of K-Means Clustering
 - ✓ Cannot cope with different densities





- Limitations of K-Means Clustering
 - √ Cannot cope with non-globular shapes

