

Unsupervised Learning: Clustering

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Clustering: An Overview

- Clustering algorithms can be classified into different categories, based on the following criteria:
 - ▶ Whether each point is assigned to exactly one cluster or several clusters with certain probabilities that add up to 1:
 - **Hard**
 - **Soft**
 - ▶ Whether all clusters are on the same level or several clusters are built in a hierarchical way:
 - **Partitional**
 - **Hierarchical**

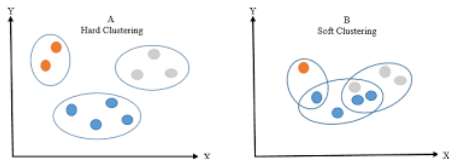


Figure: Hard vs Soft [1].

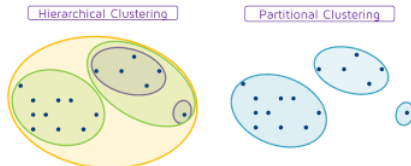


Figure: Partitional vs Heirarchical [2].

Clustering: An Overview

- Hierarchical clustering is usually done in two different ways:
 - ▶ **Agglomerative:** This is a "bottom-up" approach, Each observation starts in its own cluster, and pairs of clusters are merged as one moves up the hierarchy.
 - ▶ **Divisive:** This is a "top-down" approach, All observations start in one cluster, and splits are performed recursively as one moves down the hierarchy.

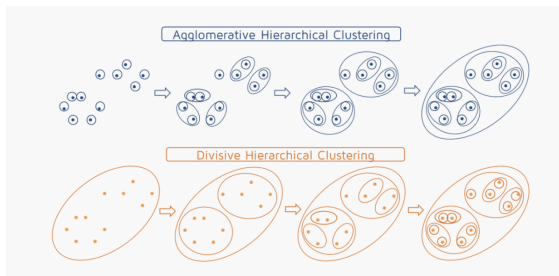


Figure: Agglomerative vs Divisive [3].

Hard Partitional Clustering: K -Means

- A particularly simple method for clustering is K -means, The idea is **to represent each cluster k by a center point \mathbf{c}_k and assign each data point \mathbf{x}_n to one of the clusters k** which can be written in terms of index sets \mathcal{C}_k
- The center points and the assignment are then chosen such that the mean squared distance between data points and center points **is minimized**:

$$J := \sum_{n=1}^N \sum_{k=1}^K r_{nk} \|\mathbf{x}_n - \mathbf{c}_k\|^2$$

- Here we introduced a corresponding binary indicator variable $r_{nk} \in \{0, 1\}$ where $k = 1, 2, \dots, K$ describing which of the K clusters the data point \mathbf{x}_n is assigned to, so that if data point \mathbf{x}_n is assigned to cluster k then $r_{nk} = 1$, and $r_{nj} = 0$ for $j \neq k$
- Now, Our goal is to find values for the $\{r_{nk}\}$ and the $\{\mathbf{c}_k\}$ so as to minimize J . we can do this through an **Iterative Procedure**

Hard Partitional Clustering: K -Means

■ To minimize J through iterating, we have to do the following algorithm :

- ① **Initialize \mathbf{c}_k with Random Value** for all $k = 1, 2, \dots, K$, It could be chosen from data values either.
- ② Minimize J with respect to r_{nk} , keeping the c_k fixed. because J is a linear function of r_{nk} this optimization can be performed easily to give a closed form solution:

$$r_{nk} = \begin{cases} 1 & \text{if } k = \operatorname{argmin}_j \|\mathbf{x}_n - \mathbf{c}_j\|^2 \\ 0 & \text{otherwise} \end{cases}$$

- ③ Minimize J with respect to \mathbf{c}_k , keeping the r_{nk} fixed. if the assignment is fixed, it is easy to show that the optimal choice of the center positions is given by:

$$\mathbf{c}_k = \frac{\sum_n r_{nk} \mathbf{x}_n}{\sum_n r_{nk}}$$

- ④ Check the convergence criteria, otherwise go to step 2.

Hard Partitional Clustering: K -Means

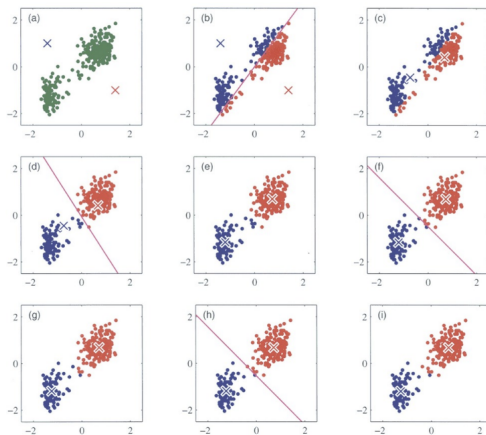


Figure: Illustration of K -Means with $K = 2$

Hard Partitional Clustering: K -Means

- Note that the result of the algorithm is **not necessarily a global optimum** of the objective function J
- It is therefore advisable to **run the algorithm several times** with different initial center locations and **pick the best result**.
- A drawback of this and many other clustering algorithms is that **the number of clusters is not determined**.
- One has to decide on a proper K in advance, or one simply runs the algorithm with several different K -values and picks the best according to some criterion.

References

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Thank You!

Any Question?