

- K-means clustering
 - Problem formulation
 - Lloyd's algorithm
 - Demo

• Review: goal of K-means algorithm is to cluster $\{\underline{x}_i\}_{i=1}^N$ into K clusters $\mathcal{C} = \{S_1, \dots, S_K\}$ such that the intra-class variance is minimized. Recall, $\underline{x}_i \in \mathbb{R}^d$.

Soln: This is an NP hard problem. We will look at an approximate soln. specifically, the Lloyd's algorithm.

1. Given, $K, \{\underline{x}_i\}_{i=1}^N, \epsilon$
2. Initialize centroids $\mathcal{C}^{(0)} = \{\underline{\mu}_1^{(0)}, \dots, \underline{\mu}_K^{(0)}\}$; $\underline{j} = 0$ iteration
3. Assign \underline{x}_n to cluster 'i' if $\underline{x}_n \rightarrow S_i^{(j)}$ [For all n]

$$i = \arg \min_{k \in \{1, \dots, K\}} \|\underline{x}_n - \underline{\mu}_k^{(j)}\|_2^2 ; d(\underline{x}_n, \underline{\mu}_i)$$
4. Update centroids:
$$\underline{\mu}_k^{(j+1)} = \frac{1}{N_k} \sum_{i \in S_k^{(j)}} \underline{x}_i$$
5. Check for stopping condition.

$$Y \left[\left(\sum_{k=1}^K \|\underline{\mu}_k^{(j+1)} - \underline{\mu}_k^{(j)}\|_2^2 \right) < \epsilon \right] \text{ output centroids and clusters}$$

threshold

Else, goto step 3.