

① Objective of k-means:

Given d_1, d_2, \dots, d_N documents, we want to find a partition of $C = \{C_1, C_2, \dots, C_k\}$, such that

$$\arg \min_C \sum_{j=1}^k \sum_{i \in C_j} \|x_i - \mu_j\|^2$$

② Purity for clustering:

N : total # of documents

k : total # of clusters

N_i : # of documents in cluster i

m_{ij} : # of instances in cluster i that belong to class j . | golden

$$\text{purity} = \frac{1}{N} \sum_{i=1}^k \left\{ \max_j m_{ij} \right\}$$

Alternative definition:

clusters $C_1 \ C_2 \ \dots \ C_k$

golden classes $g_1 \ g_2 \ \dots \ g_t$

$$\text{purity} = \frac{1}{N} \sum_{i=1}^k \left\{ \max_{j=1}^t |C_i \cap g_j| \right\}$$

③ Mixture model:

$p(x)$: prob. model for text data

$p(x)$ is mixture model \rightarrow means a mixture of uni-model distribution

Gaussian mixture model: k -Gaussian components

$$\begin{aligned} p(x_n) &= \sum_k p(x_n | z_n=k) \cdot p(z_n=k) \\ &= \sum_k \mathcal{N}(x_n | \mu_k, \Sigma_k) \cdot \pi_k \end{aligned}$$

This probability model describes how each data point x_n can be generated:

- Step 1: flip k -sided die, with prob. π_k for k -th side to select cluster c .
- Step 2: generate the values of the data point from $N(\mu_c, \Sigma_c)$.

para: $\theta = \{ \mu_k, \Sigma_k, \pi_k, k=1, 2, \dots, k \}$.

$$\theta_i = \{ \mu_i, \Sigma_i, \pi_i \}.$$

x_i : observed sample data

z_i : $\{z_i^1, \dots, z_i^k\}$. unobserved cluster labels.

$$z_i^j \in \{0, 1\}.$$

④. Submodular function:

$$\mathcal{d} = \{s_1, s_2, \dots, s_n\}.$$

$$F: 2^n \rightarrow \mathbb{R}.$$

Let $V = [n]$, $A \subseteq B \subseteq V \setminus v$.

$$F(A \cup v) - F(A) \geq F(B \cup v) - F(B) \text{ For } \forall A, B.$$

diminishing return.