

$\underline{\Omega_1}$  'N' data points , 'K' components / clusters.

initialization :- pick 'K' centroids from dataset :  
 $\mu_1, \mu_2, \dots, \mu_K$

OBJECTIVE :-  $\min \sum \text{dist}(x_i, \mu_k) \cdot \pi_{ik}$

$$= \min \sum \|x_i - \mu_k\|^2 \cdot \pi_{ik} \quad (\text{for Euclidean distance})$$

Algorithm :-

-LOOP (until convergence  $\Leftrightarrow$  no change)

E step - Compute membership  $\pi_{ik} = \begin{cases} 1 & \text{if } x_i \in \text{cluster } k \\ 0 & \text{otherwise} \end{cases}$   
(given  $\mu$ )

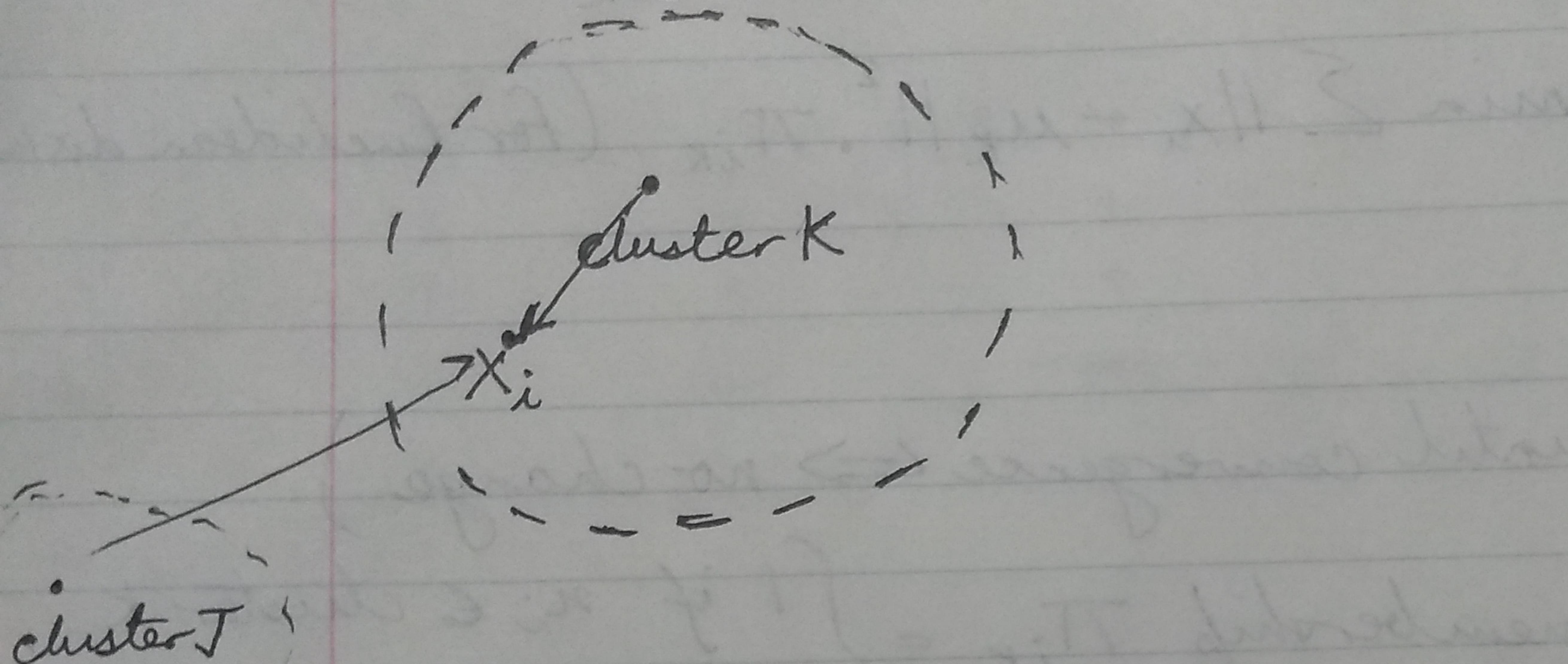
$$\pi_{ik} = 1 \Leftrightarrow k = \operatorname{arg\min}_K \text{dist}(x_i, \mu_k)$$

M step - Compute centroids  $\mu_k = \operatorname{avg}(x_i \in \text{cluster } k)$   
(given  $\pi$ )

$$= \frac{\sum_i x_i \cdot \pi_{ik}}{\sum_i \pi_{ik}}$$

A) To obtain the minimum <sup>global</sup> objective, in E-step, the only way to achieve it is ~~is~~ if the membership  $\pi_{ik}$  satisfies the equation:

$$\pi_{ik} = \begin{cases} 1 & \text{if } n_i \in \text{cluster } k \\ 0 & \text{otherwise.} \end{cases}$$



cluster J

This is because, if  $x_i$  belongs to or is a member of any other cluster, ~~distance~~ ~~distance~~ say cluster  $J$ , then, distance  $(x_i, \mu_j)$  will not be minimum and hence, objective function will not be minimum.

B) Differentiating  $\sum \|x_i - \mu_k\|^2 \cdot \pi_{ik}$  w.r.t  $\mu_k$ ,

$$\frac{d}{d\mu_k} \left( \frac{x_i \cdot \pi_{ik}}{\sum \pi_{ik}} \right) = 0, \text{ so this will be a}$$

minimum, and hence  $\mu_k^* = \arg \min_{\mu_k} (\pi_{ik} \neq 0)$  will achieve minimum objective.

- c) There are at most  $K^N$  ways to partition  $N$  datapoints into  $K$  clusters. This is a large but a finite number.  
for each iteration, we produce a new clustering based only on the previous old clustering.
- 1) If old clustering is same as the new, next clustering will again be same.
  - 2) If new clustering is different from old clustering then newer one has lower cost.
- Since the algorithm iterates a function whose domain is a finite set, the iteration must eventually enter a cycle. The cycle cannot have length  $> 1$ , because otherwise by (2) you would have some clustering which has lower cost than itself, which is a contradiction & hence not possible. Hence cycle must have exactly length = 1. Hence K-means converges in a finite no. of iterations.