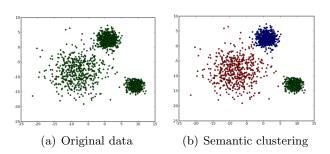
## K-means 聚类算法

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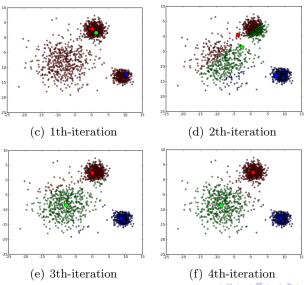
### Outline

- Motivation
- lacktriangleright K-means algorithm
- Face image clustering

### An example



### Iteration process



### **Algorithm 1** K-means algorithm

**Input**: data samples  $X = \{x_1, \dots, x_n\}$  and the number of clusters k.

Output: Clustering results of all samples.

- 1: Randomly generate k cluster centroids  $\mu_1, \dots, \mu_k$ ;
- 2: repeat
- 3: Assign each sample to the cluster having smallest distance to the corresponding cluster centeroid;
- 4: Update cluster centroids with the means of clusters;
- 5: until convergence
- 6: return Clustering result.

### **Algorithm 2** K-means algorithm

**Input**: data samples  $X = \{x_1, \dots, x_n\}$  where  $x_i \in \mathbb{R}^d$  and the number of clusters k.

Output: Clustering results of all samples.

- 1: Initialize k cluster centroids  $\mu_1, \dots, \mu_k$  where  $\mu_i \in \mathbb{R}^d$ ;
- 2: Initialize clustering indicator array  $cluster \in \mathbb{R}^n$ ;
- 3: repeat
- 4: Set  $cluster_i = \arg\min_j ||x_i \mu_j||$ ;

5: Update 
$$\mu_j = \frac{\sum\limits_{i=1}^{n} 1(cluster_i = j)x_i}{\sum\limits_{i=1}^{n} 1(cluster_i = j)};$$

- 6: until convergence
- 7: return cluster.

### Question

- Objective function?
- Initialization sensitivity?
- Convergence?
- Performance evaluation?

Objective function of K-means clustering

$$J = \sum_{i=1}^{n} \sum_{j=1}^{k} r_{ij} ||x_i - \mu_j||$$

where  $r_{ij} = 1$  if  $x_i$  is assigned to cluster j; otherwise,  $r_{ij} = 0$ . 计算可得

$$\mu_j = \frac{\sum_{i=1}^n r_{ij} x_i}{\sum_{i=1}^n r_{ij}}.$$

## Clustering accuracy

Denote  $p_i$  and  $q_i$  be the obtained clustering label and the label provided by the dataset, respectively. The ACC is defined as follows

$$ACC = \frac{\sum_{i=1}^{n} \delta(p_i, map(q_i))}{n} \tag{1}$$

where  $\delta(a,b) = 1$  if a = b; otherwise  $\delta(a,b) = 0$ .  $map(\bullet)$  is the best permutation mapping function that matches the obtained clustering label to the equivalent label of the dataset.

### Normalized mutual information

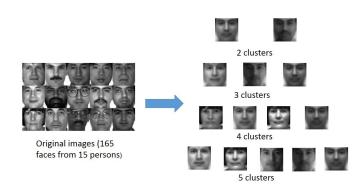
Given two random variables P and Q, NMI of P and Q is defined as

$$NMI(P,Q) = \frac{I(P;Q)}{\sqrt{H(P)H(Q)}}$$
 (2)

where I(P;Q) is the mutual information of P and Q, and H(P) and H(Q) are the entropies of P and Q, respectively. Here, the clustering results  $\widetilde{C} = \{\widetilde{C}_i\}_{i=1}^{\widetilde{c}}$  of and the ground truth labels  $C = \{C_j\}_{j=1}^c$  of all samples are viewed as two discrete random variables. NMI is specified as

$$NMI(C,\widetilde{C}) = \frac{\sum_{i=1}^{\widetilde{c}} \sum_{j=1}^{c} |\widetilde{C}_{i} \cap C_{j}| log \frac{n|\widetilde{C}_{i} \cap C_{j}|}{|\widetilde{C}_{i}||C_{j}|}}{\sqrt{(\sum_{i=1}^{\widetilde{c}} |\widetilde{C}_{i}| log \frac{|\widetilde{C}_{i}|}{n})(\sum_{j=1}^{c} |C_{j}| log \frac{|C_{j}|}{n})}}.$$
 (3)

## K-means clustering result from face images



# Thanks!