

# CS150 Discussion 12

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## K-means Clustering

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## K-means Clustering

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### The optimization problem

#### notations

- Data points:  $\mathbf{x}_i \in \mathbb{R}^D, i = 1, 2, \dots, N$
- $K$ : number of clusters
- Centers of clusters:  $\boldsymbol{\mu}_k, k = 1, 2, \dots, K$
- indicator variables:  $r_{nk} = \mathbb{I}[\mathbf{x}_i \text{ is assigned to the } k^{th} \text{ cluster}] \in \{0, 1\}$

#### goal

Find an assignment of data points to clusters, as well as a set of vectors  $\{\boldsymbol{\mu}_k\}$ , such that the sum of the squares of the distances of each data point to its closest vector  $\boldsymbol{\mu}_k$  is minimum.

## objective function

$$J = \sum_{i=1}^N \sum_{k=1}^K r_{ik} \|\mathbf{x}_i - \boldsymbol{\mu}_k\|_2^2$$

## the iterative procedure

0. choose initial values for the  $\boldsymbol{\mu}_k$
  1. minimize  $J$  w.r.t.  $r_{ik}, \forall i, k$ , keeping the  $\boldsymbol{\mu}_k$  fixed
    - $J$  involving different  $i = 1, 2, \dots, N$  are independent, so we can optimize for each  $i$  separately
      - choose  $r_{ik}$  to be 1 for whichever value of  $k$  gives the minimum value of  $\|\mathbf{x}_i - \boldsymbol{\mu}_k\|_2^2$
      - in other words, we simply assign the  $i^{th}$  data point to the closest cluster center
      - $r_{ik} = \begin{cases} 1 & \text{if } k = \arg \min_j \|\mathbf{x}_i - \boldsymbol{\mu}_j\|^2 \\ 0 & \text{otherwise.} \end{cases}$
  2. minimize  $J$  w.r.t.  $\boldsymbol{\mu}_k$ , keeping the  $r_{ik}$  fixed
    - $J$  is convex w.r.t.  $\boldsymbol{\mu}_k, \forall k = 1, 2, \dots, K$ 
      - $\frac{\partial J}{\partial \boldsymbol{\mu}_k} = 2 \sum_{i=1}^N r_{ik} (\mathbf{x}_i - \boldsymbol{\mu}_k)$
      - By setting, the gradient to 0,  $\boldsymbol{\mu}_k = \frac{\sum_i r_{ik} \mathbf{x}_i}{\sum_i r_{ik}}$
- repeat 1, 2 until convergence

## Proof of convergence

### convergence of the objective function

- $J \geq 0$  because it is a sum of squares
- Every iteration,  $J$  is strictly decreasing before the convergence
  - when  $\boldsymbol{\mu}_k$ 's are fixed, optimize over  $r_{ik}$  will reduce the value of  $J$
  - when  $r_{ik}$  fixed, optimize over  $\boldsymbol{\mu}_k$  will reduce the value of  $J$

- By the monotone convergence theorem, we know  $J$  will converge

## convergence of the $\mu_k$ 's

- Because  $N$  and  $K$  are limited, the number of the partitions is limited
  - $J$  has only limited number of values
  - $\exists T$ , when the number of iterations  $> T$ ,  $J$  is a constant  $\Rightarrow$  any  $\mu_k$  will not change