

LECTURE 28 - K-MEANS & FCM

1. K-MEANS CLUSTERING

Algorithm 1 K-Means Algorithm

```
1: Set number of clusters, M
2: Initialize cluster centers
3: repeat
4:   for i = 1 to N do
5:     Determine the closest representative,  $\Theta_j$ , for  $\mathbf{x}_i$ 
6:     Set label for data point  $i$  to  $j$ 
7:   end for
8:   for j = 1 to M do
9:     Update cluster representative  $\Theta_j$  to the mean of the points with cluster label  $j$ 
10:  end for
11: until Change in cluster centers is small
```

(1) *Run code example*

(2) Objective Function for K-means:

$$(1) \quad J(\Theta, U) = \sum_{i=1}^N \sum_{j=1}^m u_{ij} \|\mathbf{x}_i - \theta_j\|^2$$

- *How would you optimize the objective? Does the pseudo-code above match the obj. func. optimization?*
- *Does the K-means algorithm find the globally optimal solution (i.e., the cluster centers and assignments that globally minimize the objective function)?*
- *Does the K-means algorithm make any assumptions on cluster shape?*
- *Given a data set with an unknown number of clusters, can you come up with a strategy for determining the “right” number of clusters?*

2. FUZZY C-MEANS CLUSTERING

Motivation for the fuzzy memberships examples: point mid-way between clusters, overlapping clusters

So, we want soft clustering (as opposed to the hard/crisp clustering given by sequential or k-means). We can do that by defining an objective function with membership values.

The FCM objective function is:

$$J = \sum_{i=1}^N \sum_{j=1}^M u_{ij}^m d(\mathbf{x}_i, C_j) \quad \text{s.t. } 0 \leq u_{ij} \leq 1; \sum_{j=1}^M u_{ij} = 1$$

Suppose $d(\mathbf{x}_i, C_j) = \|\mathbf{x}_i - \theta_j\|_2^2$ where θ_j is the mean for cluster j , how would you minimize this objective? How do you incorporate the constraints on the membership values?

(1) *Solve for the update equation for the cluster representatives:*

$$(2) \quad J = \sum_{i=1}^N \sum_{j=1}^M u_{ij}^m \|\mathbf{x}_i - \theta_j\|_2^2$$

$$(3) \quad \frac{\partial J}{\partial \theta_j} = \sum_{i=1}^N -2u_{ij}^m (\mathbf{x}_i - \theta_j)$$

$$(4) \quad 0 = \sum_{i=1}^N (-2u_{ij}^m \mathbf{x}_i + 2u_{ij}^m \theta_j)$$

$$(5) \quad \sum_{i=1}^N u_{ij}^m \mathbf{x}_i = \theta_j \sum_{i=1}^N u_{ij}^m$$

$$(6) \quad \theta_j = \frac{\sum_{i=1}^N u_{ij}^m \mathbf{x}_i}{\sum_{i=1}^N u_{ij}^m}$$

(2) Solve for the update equation for the membership values:

$$(7) \quad L = \sum_{i=1}^N \sum_{j=1}^M u_{ij}^m \|\mathbf{x}_i - \theta_j\|_2^2 - \sum_{i=1}^N \lambda_i \left(\sum_{j=1}^M u_{ij} - 1 \right)$$

$$(8) \quad \frac{\partial J}{\partial u_{ij}} = m u_{ij}^{(m-1)} \|\mathbf{x}_i - \theta_j\|_2^2 - \lambda_i = 0$$

$$(9) \quad u_{ij} = \left(\frac{\lambda_i}{m \|\mathbf{x}_i - \theta_j\|_2^2} \right)^{\frac{1}{(m-1)}}$$

$$(10) \quad \sum_{j=1}^M u_{ij} = 1$$

$$(11) \quad 1 = \sum_{j=1}^M \left(\frac{\lambda_i^{\frac{1}{(m-1)}}}{\left(m \|\mathbf{x}_i - \theta_j\|_2^2 \right)^{\frac{1}{(m-1)}}} \right)$$

$$(12) \quad 1 = \lambda_i^{\frac{1}{(m-1)}} \sum_{j=1}^M \left(\frac{1}{\left(m \|\mathbf{x}_i - \theta_j\|_2^2 \right)^{\frac{1}{(m-1)}}} \right)$$

$$(13) \quad \lambda_i^{\frac{1}{(m-1)}} = \left(\sum_{k=1}^M \left(\frac{1}{\left(m \|\mathbf{x}_i - \theta_k\|_2^2 \right)^{\frac{1}{(m-1)}}} \right) \right)^{-1}$$

$$(14) \quad u_{ij} = \left(\sum_{k=1}^M \left(\frac{1}{\left(m \|\mathbf{x}_i - \theta_k\|_2^2 \right)^{\frac{1}{(m-1)}}} \right) \right)^{-1} \left(\frac{1}{m \|\mathbf{x}_i - \theta_j\|_2^2} \right)^{\frac{1}{(m-1)}}$$

$$(15) \quad u_{ij} = \frac{1}{\sum_{k=1}^M \frac{1}{\left(m \|\mathbf{x}_i - \theta_k\|_2^2 \right)^{\frac{1}{(m-1)}}}} \left(\frac{1}{m \|\mathbf{x}_i - \theta_j\|_2^2} \right)^{\frac{1}{(m-1)}}$$

$$(16) \quad u_{ij} = \frac{1}{\sum_{k=1}^M \left(\frac{\|\mathbf{x}_i - \theta_j\|_2^2}{\|\mathbf{x}_i - \theta_k\|_2^2} \right)^{\frac{1}{m-1}}}$$

- (1) What constraints are there on the value of the fuzzifier m ? Why?
- (2) How do results change as m approaches 1? 2? 3? How do results change with changes in m ? Can m get too big? too small?
- (3) Does FCM find the globally optimal solution? Why or why not? What sort of optimization strategy is being used in the FCM algorithm?

Algorithm 2 Fuzzy C-Means Algorithm

```

1: Set number of clusters, M, and q
2: Initialize cluster centers
3: repeat
4:   for i = 1 to N do
5:     for j = 1 to M do
6:       if  $d(\mathbf{x}_i, C_j) \neq 0$  then
7:          $u_{ij} \leftarrow \frac{1}{\sum_{k=1}^M \left( \frac{d(\mathbf{x}_i, C_j)}{d(\mathbf{x}_i, C_k)} \right)^{\frac{1}{q}}}$ 
8:       else
9:          $u_{ij} = 1$ 
10:      end if
11:    end for
12:  end for
13:  for j = 1 to M do
14:    Update cluster representatives
15:  end for
16: until Change in cluster centers is small
  
```

- (4) *What FCM membership values would a point that is located exactly halfway between two cluster centers (when clustering into two clusters) obtain?*
- (5) *What FCM membership values would an outlier data point receive? An outlier data point would be one that is far from all cluster representatives.*
- (6) *What may be some advantages of FCM over K-Means? disadvantages?*