Related Topic: Fuzzy Clustering

Input: *n* sites: $S = \{s_1, s_2, ..., s_n\}$, and a constant $m \ (m > 1)$

Output: Locations of k centers: $C = \{c_1, c_2, ..., c_k\}$

Membership of s_i to c_j : μ_{ij} (i = 1, ..., n; j = 1, ..., k)

Objective: Minimize the weighted total distance from each site to each center.

Minimize
$$\sum_{i=1}^{n} \sum_{j=1}^{k} (\mu_{ij})^{m} dist(s_{i}, c_{j})^{2}$$

where

$$0 \le \mu_{ii} \le 1, i = 1, 2, ..., n; j = 1, 2, ..., k$$

$$\sum_{j=1}^{k} \mu_{ij} = 1, \ i = 1, 2, ..., n$$

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sitecenter

where

$$0 \le \mu_{ij} \le 1, i = 1, 2, ..., n; j = 1, 2, ..., k$$

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fuzzy c-means

Minimize
$$\sum_{i=1}^{n} \sum_{j=1}^{k} (\mu_{ij})^{m} dist(s_{i}, c_{j})^{2}$$

$$0 \le \mu_{ij} \le 1, i = 1, 2, ..., n; j = 1, 2, ..., k$$

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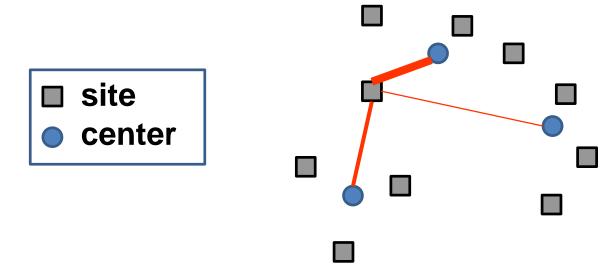
k-means

Minimize
$$\sum_{i=1}^{n} \sum_{j=1}^{k} (\mu_{ij})^{1} dist(s_{i}, c_{j})^{2}$$

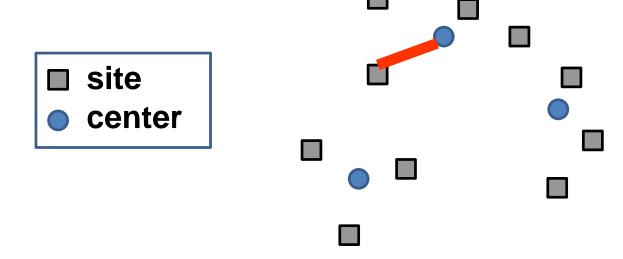
$$\mu_{ij} = 0$$
 or 1, $i = 1, 2, ..., n$; $j = 1, 2, ..., k$

$$\sum_{i=1}^{\kappa} \mu_{ij} = 1, \ i = 1, 2, ..., n$$

fuzzy c-means



k-means



Fuzzy c-means Algorithm: Iterate the following two steps from randomly specified values of μ_{ii}

(i)
$$c_j = \frac{\sum_{i=1}^{n} (\mu_{ij})^m s_i}{\sum_{i=1}^{n} (\mu_{ij})^m}, \quad j = 1, 2, ..., k$$

(ii)
$$\mu_{ij} = \left[\sum_{h=1}^{k} \left(\frac{dist(s_i, c_j)}{dist(s_i, c_h)}\right)^{\frac{2}{m-1}}\right]^{-1}$$
 for all i and j

$$\mu_{ij} = \left[\sum_{h=1}^{k} \left(\frac{dist(s_i, c_j)}{dist(s_i, c_h)}\right)^{\frac{2}{m-1}}\right]^{-1}$$
 for all i and j

When
$$m = \infty$$

$$\mu_{ij} = \sum_{h=1}^{k} \left(\frac{dist(s_i, c_j)}{dist(s_i, c_h)} \right)^0 = 1/k$$

When
$$m => 1$$

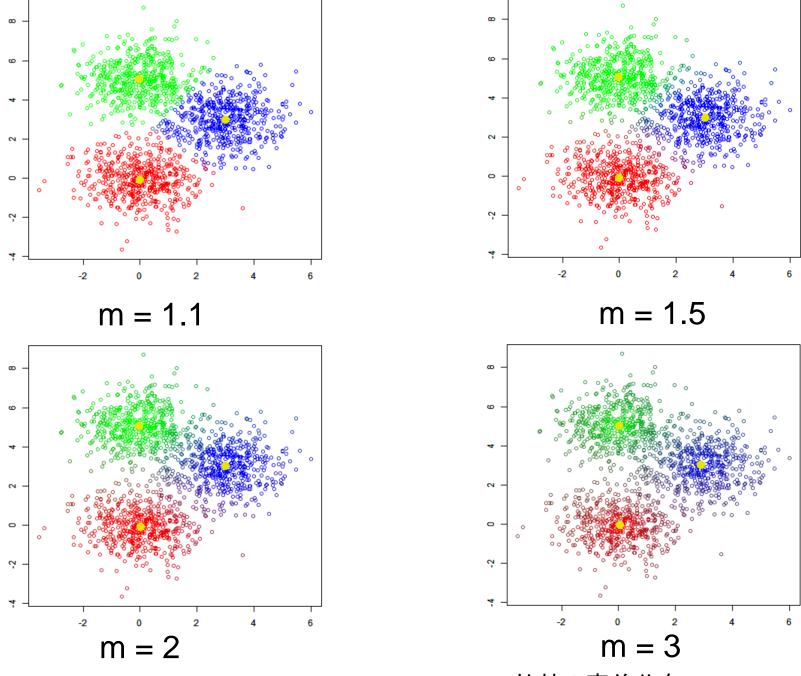
$$\mu_{ij} ==> \left[\sum_{h=1}^{k} \left(\frac{dist(s_i, c_j)}{dist(s_i, c_h)} \right)^{\infty} \right]^{-1} = 0 \text{ or } 1$$

Exercise 4-1:

Clearly demonstrate the effects of m on the clustering results by the fuzzy c-means algorithm through computational experiments on a test data set (i.e., create a test data set which can be used for clearly demonstrating the effects of m).

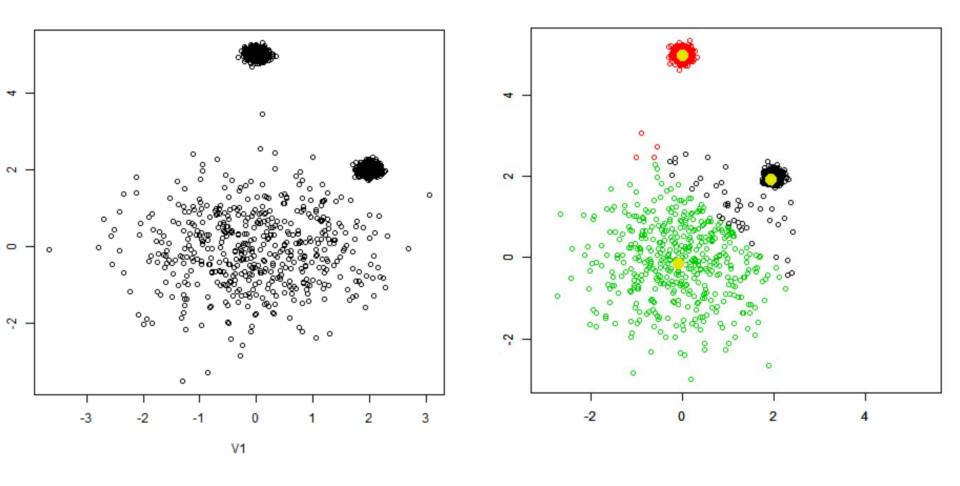
Exercise 4-2:

Clearly demonstrate the difference between the *k*-means algorithm and the fuzzy c-means algorithm through computational experiments on a test data set (i.e., create a test data set which can be used for clearly demonstrating the difference between the *k*-means algorithm and the fuzzy c-means algorithm).

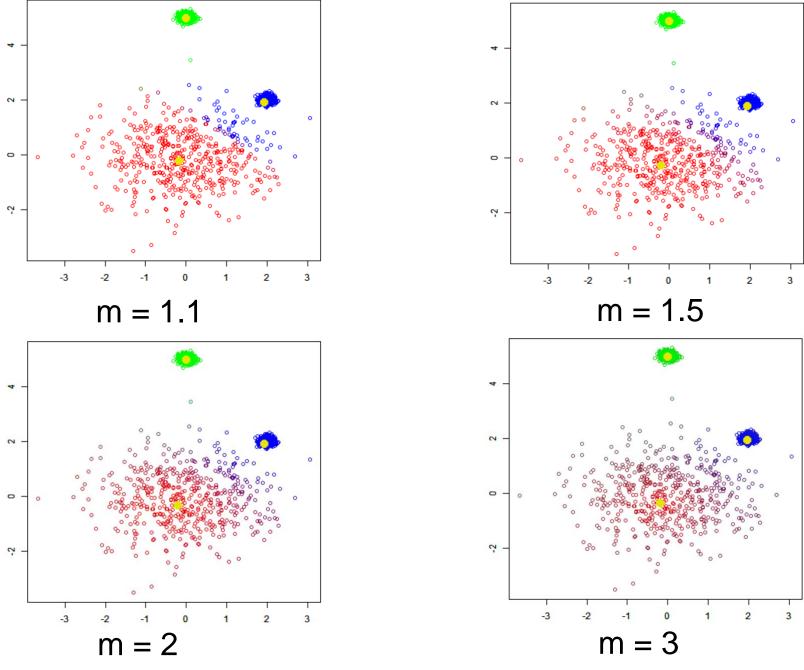


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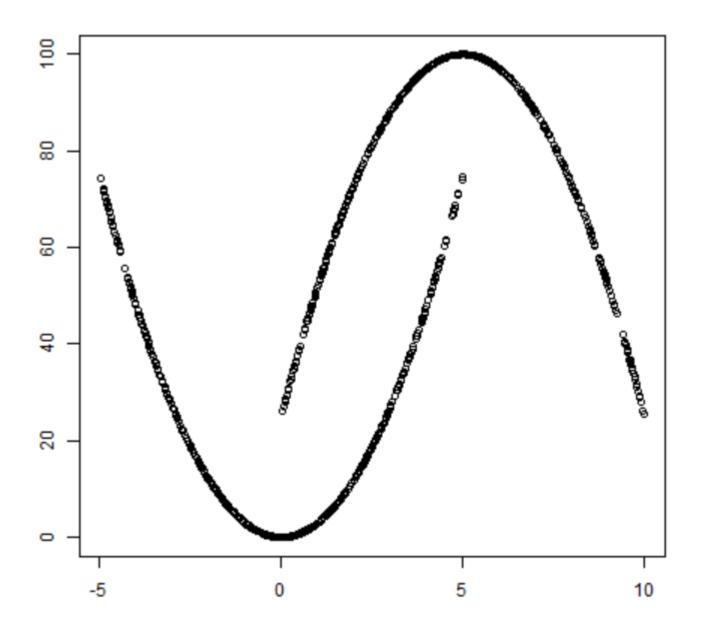
k-means



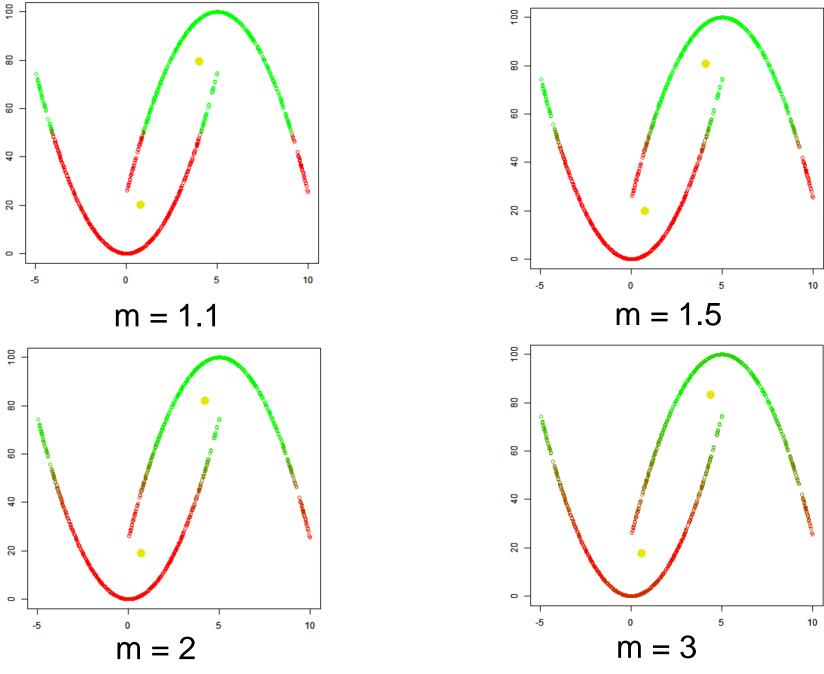
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Use of a user-defined hyper parameter (*m* in fuzzy c-means)

Positive Aspects:

A more desirable result can be obtained by appropriately specifying the value of m (than the case of the fixed value of m)

Different results can be obtained by examining different values of m (we can choose one of them based on our preference).

Negative Aspects:

It is not always easy to appropriately specify the value of m.

Undesirable results can be obtained when the value of *m* is inappropriate.

Example: If m is specified as m = 1, the algorithm does not work.

(i)
$$c_j = \frac{\sum_{i=1}^{n} (\mu_{ij})^m s_i}{\sum_{i=1}^{n} (\mu_{ij})^m}, \quad j = 1, 2, ..., k$$

(ii)
$$\mu_{ij} = \left[\sum_{h=1}^{k} \left(\frac{dist(s_i, c_j)}{dist(s_i, c_h)}\right)^{\frac{2}{m-1}}\right]^{-1}$$
 for all i and j