

MODIFIED FUZZY C-MEANS ALGORITHM IN MEDICAL IMAGES

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Abstract- Image segmentation is a common area of image processing. The research work done in this paper is based on image segmentation using improved fuzzy c-mean (FCM) algorithm. The improved fuzzy c-mean algorithm is composed by modifying the distance measurement of the original fuzzy c-mean algorithm. The original Euclidean distance in the fuzzy c-mean algorithm is replaced by correlation distance. As a result the new improved fuzzy c-means algorithm is obtained. The acquired algorithm is more robust when associated with core fuzzy c-mean algorithm. The exploration done on the real MR images exhibits that the proposed algorithm has enhanced performance.

Keywords-fuzzy c means, clustering, MRI

1. Introduction

Image segmentation is a stimulating task in image scrutiny. There huge number of methods available in image segmentation process. The fuzzy c-mean procedure that has been effectively applied to analysis, clustering of data points in the field of industries, astronomy, geology, medical image, target recognition, image segmentation, pattern recognition. An image can be appearing in different feature spaces, and fuzzy c-mean method classifies by grouping the similar data points in the feature space into clusters. Image segmentation plays important role in medical image. In the field of medical diagnosis an extensive diversity of imaging techniques is presently available, such as radiography, computed tomography (CT) and Magnetic resonance imaging (MRI) [1],[2]. In the recent times, magnetic resonance image is the most effectively used for diagnostic image examination for brain diagnostic image examination for brain diseases such as tumor. Even through original fuzzy c-mean algorithm yields good results for segmenting noise free images, it fail to segment image corrupted by noise, outliers and other imaging artifacts.

Medical image segmentation is an essential step for most successive image analysis task. This paper presents an image segmentation approach using improved fuzzy c-mean algorithm [3], [4].

2. Data Mining

Data mining, *the extraction of hidden predictive information from large databases*, is a powerful new technology with great potential to help companies focus on the most important information in their data warehouses.

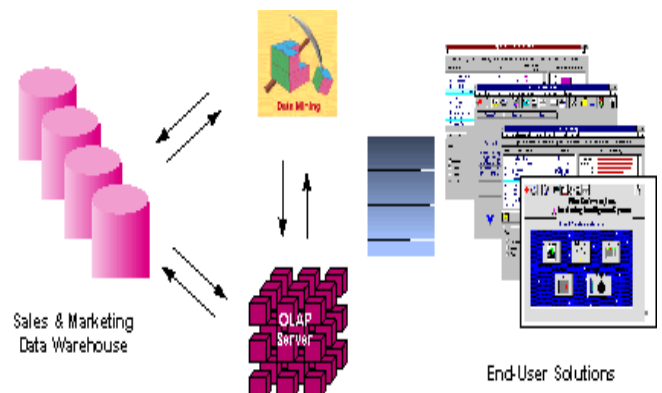


Fig1: Data Mining Architecture

3. Clustering

The process of grouping a set of physical or abstract objects into classes of similar objects is called clustering. A cluster is a collection of data objects that are similar to one another within the same cluster and are dissimilar to the objects in other clusters. There are two properties in clustering:

- Homogeneity inside clusters: the data, which belong to one cluster, should be as similar as possible.
- Heterogeneity between the clusters: the data, which belongs to different clusters, should be as different as possible.

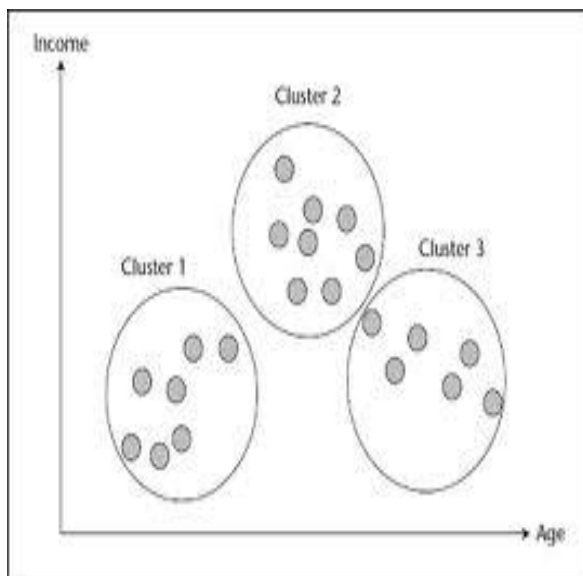


Fig2: Clustering Example

4. Definition of clustering

Clustering has been defined in many forms by their by various authors.

Clustering is “the process of organizing objects into groups whose members are similar in some”.

Clustering is an unsupervised data mining technique, which groups the input into K regions based on some similarity/dissimilarity metric. The main objective of any clustering technique is to produce a K n

partition matrix $U(X)$ of the given data set X , consisting of n patterns,
 $X = x_1, x_2, \dots, x_n$.

5. Improved fuzzy c-mean algorithm

Fuzzy c-mean (FCM) algorithm, also known as fuzzy ISODATA, was introduced by Bezdek [6] as an extension to Dunn's algorithm [7]. The fuzzy c-mean based algorithms are the most commonly used fuzzy clustering algorithms in practice. Let, $X = x_1, x_2, \dots, x_N$ where $x_i \in R^n$ present a given set of feature data.

The main aim of fuzzy c-mean algorithm is to minimize the fuzzy c-mean cost function.

$$J(U, V) = \sum_{j=1}^C \sum_{l=1}^N \mu_{ij}^m \|x_i - v_j\|^2$$

$$V = v_1, v_2, \dots, v_c$$

are the cluster centers.

$U = \mu_{ij} \in [0, 1]$ is a fuzzy partition matrix, in which each member μ_{ij} indicates the degree of membership between the data vector x_i and the cluster J . The values of matrix U should satisfy the following conditions:

$$\mu_{ij} \in [0, 1], \forall i = 1 \dots N, \forall j = 1 \dots, C \quad 2$$

$$\sum_{j=1}^C \mu_{ij} = 1, \forall i = 1, \dots, N \quad 3$$

The exponent $m \in [1, \infty)$ is the weighting exponent, which determines the fuzziness of the clusters. The most commonly used distance norm is the Euclidean distance $d_{ij} = \|x_i - v_j\|$, although Babuska suggests that other distance norm could produce better results [8]. The Euclidean distance in improved fuzzy c-mean algorithm is replaced by the correlation distance. And this improved fuzzy c-mean algorithm is to be more robust than the original fuzzy c-mean algorithm.

Minimization of the cost function $J(U,V)$ is a nonlinear optimization problem, which can be minimized with the following iterative algorithm:

Step 1: Initialize the membership matrix U with random values so that the conditions (2) and (3) are satisfied. Choose appropriate exponent m and the termination criteria.

Step 2: Calculate the cluster centers V according to the equation:

$$v_j = \frac{\sum_{i=1}^N \mu_{ij} m_{xi}}{\sum_{i=1}^N \mu_{ij}}, \forall j = 1, \dots, c$$

Step 3: Calculate the new distance by correlation:

$$d_{ij} = \rho_{x,y} = \text{corr}x,y = \frac{\text{cov}x,y}{\sigma_x\sigma_y}$$

Step 4: Update the fuzzy partition matrix U :

If $d_{ij} > 0$ (indicating that $x_i \neq v_j$)

$$\mu_{ij} = \frac{1}{\sum_{k=1}^c \left(\frac{d_{ij}}{d_{jk}}\right)^{\frac{2}{m-1}}}$$

Else

$$\mu_{ij} = 1$$

Step 5:

If the termination criteria have been met, stop.

Else go to **Step 2**

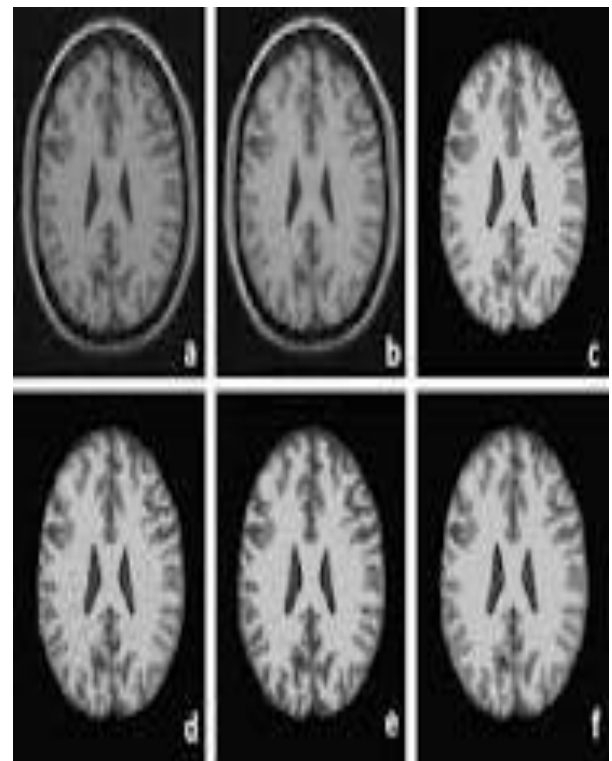
A suitable termination criterion could be to calculate the cost function (Eq. 1) and to see whether it is below a certain tolerance value or if its improvement compared to the previous iteration is below a certain threshold [9]. Also the

maximum number of iteration cycles can be used as a termination criterion. Experiments are conducted on real images to examine the performance of the proposed improved fuzzy c-mean technique in segmenting the MR-images.

6. Experimental results

The proposed improved fuzzy c-mean algorithm is implemented using MATLAB and tested on real images to explore the segmentation accuracy of the proposed approach.

The proposed approach of image segmentation using improved fuzzy c-means algorithm eliminates the effect of noise greatly.



I. Fig (a) & (b) are original images. **II.** (b)&(c) are results of core fuzzy c mean algorithm. **III.** Images (e)&(f) are results of proposed algorithm.

7. Conclusion

Images corrupted by noise, outliers and other imaging artifact. In the proposed improved fuzzy c-mean algorithm, are incorporated to control the trade-off between them. The algorithm is formulated by modifying the distance measurements of the standard fuzzy c-mean algorithm to allow the labeling of a pixel to be influenced by other pixels and to control the noise effect during segmentation. The experimental results suggested that the proposed algorithm performed well than other fuzzy c-mean extension, segmentation algorithm.

8. References

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