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Vol. 6, Issue 3, March 2017

Brain Tumor Automated Detection and Segmentation

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ABSTRACT: More precisely, we propose to use Support Vector Machines (SVM) which is one of the most popular and well motivating classification methods. The experimental study will be carried on real and simulated datasets representing different tumor shapes, locations, sizes and image intensities. Tumor is an uncontrolled growth of tissue in any part of the body. The tumor is of different types and they have different characteristics and different treatment. This paper is to implement of Simple Algorithm for detection of range and shape of tumor in brain MR Images. Normally the anatomy of the Brain can be viewed by the MRI scan or CT scan. MRI scanned image is used for the entire process. The MRI scan is more comfortable than any other scans for diagnosis. It will not affect the human body, because it doesn't practice any radiation. It is centered on the magnetic field and radio waves. There are dissimilar types of algorithm were developed for brain tumor detection. But they may have some drawback in detection and extraction. After the segmentation, which is done through k-means clustering and fuzzy c-means algorithms the brain tumor is detected and its exact location is identified. Comparing to the other algorithms the performance of fuzzy c-means plays a major role. In Proposed system we develop K-Means Algorithm and Fuzzy C-means for Brain tumor segmentation. The patient's stage is determined by SVM classifier , whether it can be cured with medicine or not By using this we get more accurate result as compare to existing system.

KEYWORDS: Tumor, MRI Scan, CT scan, K-Means clustering, Fuzzy c-means, SVM.

I. INTRODUCTION

Tumor segmentation from MRI data is an important but time-consuming and difficult task often performed manually by medical experts. Radiologists and other medical experts spend a substantial amount of time segmenting medical images. However, accurately labeling brain tumor is a very time-consuming task, and considerable variation is observed between doctors [2].

Throughout the few years, different segmentation methods have been used for tumor detection but it is time consuming process and also gives inaccurate result. So, computer aided system can be designed for accurate brain tumor detection from MRI images. Brain tumor can be broadly classified as primary brain tumor(the tumor originates in the brain) and secondary brain tumor (spread to brain from somewhere else in the body through metastasis) Primary brain tumors do not spread to other body parts and can be malignant or benign and secondary brain tumors are always malignant. Malignant tumor is more dangerous and life threatening than benign tumor. The detection of malignant tumor only for that we need to extract the exact brain tissues for that we have performed the skull removing process in that we have used the horizontal, diagonal, anti-diagonal and vertical masks to perform the erosion and dilation which is results in to the skull masked image which further proceed to segmentation. Labeling of connected components ('objects') is one of the most important tools of image processing. It is the basis for the generation of object features as Well as of some kind of filtering, *i.e.*, removing of noisy objects or holes in objects. The criteria for removing an object or a hole can be chosen



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Vol. 6, Issue 3, March 2017

extremely flexible based on the object features. The task of labeling (object filling, region detection) is to assign labels (mostly unsigned integers) to the pixels in such a way that all pixels belonging to a connected component of the image are assigned the same label, and pixels belonging to different components get different labels [5] This paper proposed concept for brain tumor segmentation and finally the detection of brain tumor. Normally the structure of the Brain can be viewed by the CT scan or MRI scan. In this paper the MRI scanned image is taken for the

whole process. The MRI scan is more comfortable and suitable than CT scan for diagnosis. It is not affect the humanbody. Because of this method doesn't use any radiation. This process based on the magnetic field and radio waves. There are different types of algorithm were developed for brain tumor detection. But they may have some drawback in extraction and detection of brain tumor. In this paper, two algorithms are used for segmentation. K-means clustering algorithm and Fuzzy-C means algorithm. So it gives the accurate result for brain tumor segmentation. Tumor is due to the uncontrolled growth of the tissues in any part of our body. The tumor stage may be primary or secondary. If it is an origin, then it is known as primary stage. If the part of the tumor is spread to another place and grown as its own then it is known as secondary stage. Normally brain tumor affects CSF (Cerebral Spinal Fluid). It causes for strokes. The physician gives the treatment for the strokes rather than the treatment for tumor. So detection of tumor is important for that further treatment. The lifetime of the person who affected by the brain tumor will increase if it is detected at current stage correctly. That will increase the lifetime about 1 to 2 years. Normally tumor cells are of two types. They are Mass and Malignant tumor. The detection of the malignant tumor is difficult to mass tumor. In this paper we focused on detection of brain tumor with the help of Brain MRI images. The developing platform for the detection is java. Because it is easy to develop and execute. At the end, we are providing systems that detect the brain tumor and its shape.

SN.	Author and Title	Proposed System	Implemented Concepts
1.	Samir Kumar Bandhyopadhyay,	This paper has proposed a system of	Implanting the K-mean algorithm
	TuhinUtsab Paul ,	image registration and data fusion	which consists of multiple phases.
	"Automatic Segmentation of	theory adapted for the segmentation	First phaseconsists of registration
	Brain Tumor from	of MR images. This system	of multiple MR images of the
	Multiple Images of Brain MRI"	provides an efficient and fast way	brain taken along adjacent layers
		for diagnosis of the brain tumor	of brain. In the second phase,
		called K-means algorithm.	these registered images are fused to
			produce high quality image for the
			segmentation. Finally,
			segmentation is done by improved
			K -means algorithm with dual
			localization methodology.
2.	A. Meena, K. Raja, " Spatial	Meena and Rajaproposed an	The proposed FCM successful able
	Fuzzy C-Means PET Image	approach of Spatial Fuzzy C-means	to join the the spatial
	Segmentation of	(PET-SFCM) clustering algorithm	neighborhood information with
	Neurodegenerative	on Positron Emission Tomography	classical FCM and updating the
	Disorder"	(PET) scan image datasets.	objective function of each cluster.
			It exploit the segmentation which
			used for quick bird view for any
			problem of K-means.
3	SumanTatiraju, Avi Mehta, "	In this project, we look at three	The segmentation technique
	Image Segmentation using k-	algorithms namely K Means	addresses the problem of
	means clustering,	clustering, Expectation	segmenting an image into different
	EM and Normalized Cuts"	Maximization and the Normalized	regions. So the we can analyze
		cuts and compare them for image	both k-mean and C-mean
		Segmentation	algorithm in easy way.

II. LITERATURE SURVEY



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4.	AjalaFunmilola,	Funmilola et al proposed the Fuzzy	The F-K-C means focused
	" Fuzzy k-c-means Clustering	K-C-means method, which carries	attention on Clustering methods.
	Algorithm for Medical Image	more of Fuzzy C-means properties	These k-mean and C-mean
	Segmentation"	than that of K-means.	algorithms were combined
			together to come up with another
			method called fuzzy k-c-means
			clustering algorithm, which has a
			better result in terms of time
~			utilization.
5.	Besniba Wilson, Julia Punitna	Wilson and Dhas used K-means and	Succentibility weighted imaging
	Analysis of Fuzzy C Moons and	detect the iron in brain using SWI	(SWI) is a nouro imaging
	K-Means Segmentation	technique	technique which uses tissue
	Algorithm for Iron Detection in	teeninque.	magnetic susceptibility differences
	Brain SWI using Matlah"		to generate a unique contrast. The
	Drain 5 111 doing Triando		extraction of the iron region in the
			brain is made by K-means and
			Fuzzy C-means clustering method.
6.	M.H. FazelZarandia,	This paper proposed a dip study of	A brief knowledge about tumor
	"Systematic image processing	brain tumor. It describes different	like glial tumor which cover 30 %
	for diagnosing brain tumors:	type of diagnosis approaches.	of all brain tumor.
	A Type-II fuzzy expert system"		
_			
7.	Samarjit Das , "Systematic	In the field of pattern recognition	The managed frame
	image processing for	alle to the fundamental involvement	tashriqua Euglidean distance has
	A Type II fuzzy expert system	of numan perception and inadaguagy of standard	been used to obtain the
	approach"	Mathematics to deal with its	membership values of the objects
	approach	complex and ambiguously defined	in different clusters: in our present
		system, different fuzzy techniques	work along with Euclidean
		have been applied as an appropriate	distance we have used other
		alternative	distances like Canberra distance,
			Hamming distance to see the
			differences in outputs.
8.	Vignesh Rajesh, " brain	This paper has suggested a	The implemented method
	tumor segmentation and its	synergistic and an effective	enhance the quality of the tumor
	area calculation in brain mri	algorithm for the detection of brain	images acquired by the aid
	images	tumors based on Median filtering, K	of MRI and then to detect the size
	using k-mean clustering and	Means Segmentation, FCM	of the tumors, approximate
	luzzy cmean	Segmentation, and finally, threshold	reasoning are applied.
9	Krishna Kant Singh " A Study	In this paper the author gives a	Implementation of segmentation
	Of Image Segmentation	study of the various	technique those are color-based-
	Algorithms For	algorithms that are available for	segmentation, pixel-based
	Different Types Of Images"	color images, text and	segmentation and edge-based
		gray scale images.	segmentation.
10	payalmistry, shagunakhauri,	In this paper proposed k-means and	K-Means and Fuzzy C- Means In
	sayalipatil, s.p.tondare, "	C-mean to extract the features from	this process the tumor is extracted
	segmentation of brain tumor	the images.	from the MR image and its exact
	and its area		position and the shape also



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11	calculation in brain mr images using k-mean clustering and fuzzy c- mean algorithm" D.Judehemanth, D.Selvathi, J.Anitha " Effective fuzzy Clustering algorithm for abnormal MR brain image segmentation"	In this paper, the application of modified FCM algorithm for MR brain tumor detection is explored	determined. The stage of the tumor is displayed based on the amount of area calculated from the cluster The effectiveness of thte FCM algorithm in terms of computational rate is improved by modifying the cluster center .
12	ipekoguz and milansonka "logismos-b: layered optimal graph image segmentation of multiple objects and surfaces for the brain"	This paper propose a novel algorithm, LOGISMOS-B, based on probabilistic tissue classification, generalized gradient vector flows and the LOGISMOS graph segmentation framework.	Automated reconstruction of the cortical surface is one of the most challenging problems in the analysis of human brain magnetic resonance imaging (MRI). The main goal of this manuscript is to present a novel method to obtain more accurate and robust cortical segmentations.
13	zhuowentu , xiangbai "auto- context and its application to high-level vision tasks and 3d brain image segmentation"	This paper, propose a learning algorithm, auto-context. Given a set of training images and their corresponding label maps, we first learn a classifier on local image patches.	Auto-context integrates low-level and context information by fusing a large number of low-level appearance features with context and implicit shape information. The resulting discriminative algorithm is general and easy to implement.
14	zhuowentu, katherine l. narr, piotrdollár, ivodinov, paul m. thompson, arthur w. toga "brain anatomical structure segmentation by hybrid discriminative/generative models"	In this paper, a hybrid discriminative/generative model for brain anatomical structure segmentation is proposed. The learning aspect of the approach is emphasized. In the discriminative appearance models, various cues such as intensity and curvatures are combined to locally capture the complex appearances of different anatomical structures.	A probabilistic boosting tree (PBT) framework is adopted to learn multiclass discriminative models that combine hundreds of features across different scales. The parameters to combine the discriminative appearance and generative shape models are also automatically learned. Thus, low- level and high-level information is learned and integrated in a hybrid model.
15	zexuanji, yongxia, quansen sun, qiangchen, deshenxia, daviddaganfeng "fuzzy local gaussian mixture model for brain mr image segmentation"	In this paper, we assume that the local image data within each voxel's neighborhood satisfy the Gaussian mixture model (GMM), and thus propose the fuzzy local GMM (FLGMM) algorithm for automated brain MR image segmentation.	The algorithm estimates the segmentation result that maximizes the posterior probability by minimizing an objective energy function, in which a truncated Gaussian kernel function is used to impose the spatial constraint and fuzzy memberships are employed to balance the contribution of each GMM.



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Vol. 6, Issue 3, March 2017

Table 1: Literature survey

III. MATHEMATICAL MODEL

Mathematical equation in K-means clustering

1. $M = \frac{\sum_{i:c(i)=k} Xi}{Nk}, k=1, 2, \dots, K.$ 2. D (i) = arg min||X_i- M_k||², i=1,2, \dots, N. Mathematical equation in Fuzzy-C means clustering $Y_{m} = \sum_{i=1}^{N} \sum_{j=1}^{C} M_{ij}^{m} ||X_{i} - C_{j}||^{2}$ Where,

m= any real number greater than 1,

 M_{ii} = degree of membership of X; in the cluster j,

 X_i = data measured in d-dimensional,

 R_i = d-dimension center of the cluster,

The update of membership Mij and the cluster centers R, are given by:

 $M_{ij} = \frac{1}{\sum_{K=1}^{C} \left(\frac{||Xi-Cj||}{||Xi-Cj||}\right) \frac{2}{m-1}}$

 $R_{j} \!\!=\! \frac{\sum_{i=1}^{N} \textit{Xi.Mijm}}{\sum_{i=1}^{N} \textit{Mijm}}$

Let us consider S as a system for BRAIN TUMOR SEGMENTATION AND ITS AREA CALCULATION USING K-MEANS AND C-MEANS ALOGRITHM $S = \{....\}$

INPUT:

• Identify the inputs

 $F = \{f_1, f_2, f_3, \dots, f_n\}$ 'F' as set of functions to execute commands. $\}$

I = { $i_1, i_2, i_3...$ |'I' sets of inputs to the function set }

 $O = \{o_1, o_2, o_3..., | O' \text{ Set of outputs from the function sets} \}$

 $S = \{I, F, O\}$

- $I = {$ Query submitted by the user,...}
- $O = \{Output of desired query, ...\}$
- $F = \{Functions implemented to get the output,$

K-means clustering, Fuzzy C –means clustering}

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SPACE COMPLEXITY
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The space complexity depends on Presentation and visualization of discovered patterns. More the storage of data more is the space complexity.

TIME COMPLEXITY

Check No. of patterns available in the datasets = n

If (n>1) then retrieving of information can be time consuming.

So the time complexity of this algorithm is $O(n^n)$.

 Φ = Failures and Success conditions.

Failures:

- 1. Huge database can lead to more time consumption to get the information.
- 2. Hardware failure.
- 3. Software failure.



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Vol. 6, Issue 3, March 2017

Success:

Search the required information from available in Datasets. User gets result very fast according to their needs.

IV. PROPOSED SYSTEM ARCHITECTURE



Figure 1 : System Architecture

1. Pre-processing

According to the need of the next level the pre-processing step convert the image. It performs filtering of noise and other artifacts in the image and sharpening the edges in the image. RGB to grey conversion and Reshaping also takes place here. It includes median filter for noise removal. The possibilities of arrival of noise in modern MRI scan are very less. It may arrive due to the thermal effect. The main aim of this paper is to detect and segment the tumor cells. But for the complete system it needs the process of noise removal.

2. Segmentation using K-means

Steps

- 1. Give the no of cluster value as k.
- 2. Randomly choose the k cluster centers
- 3. Calculate mean or center of the cluster
- 4. Calculate the distance b/w each pixel to each cluster center
- 5. If the distance is near to the center then move to that cluster.
- 6. Otherwise move to next cluster.
- 7. Re-estimate the center.
 - 8. Repeat the process until the center doesn't move.



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Vol. 6, Issue 3, March 2017

3. Segmentation using Fuzzy C means and SVM

The fuzzy logic is a way to processing the data by giving the partial membership value to each pixel in the image.

The membership value of the fuzzy set is ranges from 0 to 1.

Fuzzy clustering is basically a multi valued logic that allows intermediate values i.e., member of one fuzzy set can also be member of other fuzzy sets in the same image.

There is no abrupt transition between full membership and non-membership.

The membership function defines the fuzziness of an image and also to define the information contained in the image.

4. Feature Extraction

The feature extraction is extracting the cluster which shows the predicted tumor at the FCM output. The extracted cluster is given to the thresholding process.

It applies binary mask over the entire image. It makes the dark pixel become darker and white become

brighter.

In threshold coding, each transform coefficient is compared with a threshold.

If it is less than the threshold value then it is considered as zero.

If it is larger than the threshold, it will be considered as one.

The thresholding method is an adaptive method where only those coefficients whose magnitudes are above a threshold are retained within each block.

5. Approximate reasoning

In the approximate reasoning step the tumor area is calculated using the binarization method. That is the image having only two values either black or white (0 or 1).

V. EXPERIMENTAL RESULT AND DISSCUSSION

Proposed Brain tumor detection system is improve with segmentation of preprocessed image then resulted image goes with object labeling and feature extraction. Extracted features used to train SVM and the database of feature is use for pattern matching and test the system.

1. Calculate the area of tumor=

Area= $\sqrt{P} * 0.264$ Where, p= total no of white pixels of threshold image. And 0.264 is the 1 pixel size.



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Vol. 6, Issue 3, March 2017



In above figure the result of proposed system(SVM) is explained there, the first is original MRI image which is having some noise in it. The noise is removed with median filter the result of median filter is shown in (B) image. Then the morphological filtering is used on the preprocessed image and the masks are used to remove the fatty tissues and skull bone. Then the resulted image is go under the segmentation with thresholding, here we have used three colors to show regions according to the intensity slices. After that the resulted image of segmentation goes under object

labeling with HOG algorithm and also does the extraction of texture, color and shape context feature. The resulted image of segmentation and object labeling goes under SVM learning, for SVM the linear classifier algorithm is used here and another features that are Mean, Standard deviation, Entropy, RMS, Varience, Smoothness, Kurtosis, Skiwness, IDM, Contrast, Corelation, Energy, Homogenety extracted and stored on the .mat file. This all features are stored in database and that are use for making patterns of the classes that are benign tumor or malignant tumor.



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Figure 2: Result analysis of Execution time of various Segmentation methods

The above figure 2 shows the result analysis of Execution time of various Segmentation methods. In that the k-means clustering gives fine result than the other methods that's why here in improved system K-Means clustering is used for segmentation.



Figure 3: Result analysis of Execution time of Proposed system and Existing system

The proposed system gives a faire result for the input that is MRI images. The proposed method includes the k-means for segmentation HOG for segmentation and the SVM for pattern mapping and pattern matching process. Before this system there exist the other methods to identify the brain tumor one of them is FCM mean and only simple k-means is used for tumor detection. In above number of images verses accuracy graph the red line shows the graph of proposed



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Vol. 6, Issue 3, March 2017

algorithm, the green is FCM mean and the blue is k-means algorithm performance according to the accuracy towards the tumor detection.



Figure 4: Result analysis of Accuracy of Proposed system and Existing system

The proposed system is also very sensitive to the errors, because the small error will take the situation in ambiguous state which is not good for diagnosis of tumor so here we are taking a resulted graph of number of images verses overall error in system. Again same FCM mean and k means algorithms are use to compare individual performance with the proposed method and the result of all are compared and we found that the proposed system having less errors in the system.

VI.CONCLUSION

There are different types of tumors available. They may be mass in the brain or malignant over the brain. Suppose if it is a mass, then K- means algorithm is enough to extract it from the brain cells. If there is any noise present in the MR image it is removed before the K-means process. The noise free image is given as input to the k-means and tumors are extracted from the MRI image. The performance of brain tumor segmentation is evaluated based on K-means clustering. Dataset consists of Magnetic Resonance Imaging (MRI) size of 181X272. The MRI image dataset that we have utilized in image segmentation technique is taken from the publicly available sources. The brain image dataset is divided into two sets. Training dataset and testing dataset. Thus, the pre-processing is done by filtering. Segmentation is done by advanced K-means algorithm and fuzzy c means algorithm.Feature extractions is done by thresholding and finally, approximate reasoning method to recognize the tumor shape and position in MRI image using SVM. The stage of the tumor is based on the area of tumor.

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