Cerebrum Tumor Segmentation and Detection Technique for MRI Imaging

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Abstract: The cerebrum tumors are the most well-known and forceful sickness, prompting an extremely short future in their most noteworthy evaluation. Accordingly, treatment arranging is a key stage to improve the personal satisfaction of patients. Generally, various medical image modalities like Magnetic Resonance Imaging (MRI), Computed Tomography (CT) and ultrasound image are used to evaluate the cerebrum tumor in a brain, lung, liver, breast, prostate etc. MRI images are very much useful for different types of brain tumor exposure and segmentation. A plethora of methods like k-means clustering, Fuzzy C-Means, SOM clustering, Deep Convolution Neural Networks (DNN), SVM, Convolutional Neural Networks (CNN) for cerebrum brain tumor detection from MRI images. This paper concentrated on mind cerebrum tumor recognition calculations that have been planned so distant to recognize the area of the cerebrum tumor.

Keywords: Cerebrum Brain Tumor, Magnetic Resonance Imaging Images, K-means, Fuzzy C-Mean, SVM, DNN, sMorphological operation, CNN.

I. INTRODUCTION

In the present time digital image processing plays a significant responsibility in medical image field. Medical imaging is the rising and difficult ground. Medicinal imaging is advantageous in diagnosis of the different types of cancerous disease and tumors. Several people suffer from brain tumor which is a severe and risky disease that leads to death in many cases. Medical imaging provides appropriate and suitable diagnosis of brain tumor. There are lots of techniques available to detect brain tumor from MRI images. The one of the brain tumor namely by cerebrum brain tumor characterizes the abnormal expansion of infected tissues and infected cells multiplication because of this the regular example of cell development and passing is fizzle. The two stages of brain tumor are primary and secondary stage.

A primary stage of brain tumor initiates in brain. A few essential mind tumors are either amiable or harmful. Amiable malignancy does not degree from single tissue of the body organs to the nearby tissue of the body organs. It can be formed into cerebrum cells, nerve cells and organs.

An optional phase (secondary) of mind tumor, otherwise called a metastatic cerebrum cancer, occurs subsequent to cancer cells degree to the cerebrum from elective body organs, like kidney, lung, skin or bosom. Auxiliary cerebrum brain tumors are constantly threatening which can begin from the single tissue of the body organ and increase towards the other parts rapidly [1].

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In the initial stage the tumor can be expelled or considered as a optional stage, the tumor infection spreads because of this expulsion of tumor in the body occasionally remain same and develops back again so this is the most critical and serious issue in the auxiliary phase of brain tumor. The cause of this happening is due to incorrectly measure the area of tumor zone. Imaging modalities is used to discover and to quantify infected location of brain tumor is possible by-

1) MRI scan
2) CT scan
3) Ultra sound etc.

MRI has the capability to provides three types of images of the brain a) T1-weighted image b) T2-weighted image and c) Proton density images. MRI images are captured into three coordinate planes namely by axial plane, coronal plane and sagittal plane as shown in figure 1.

Fig.1. Brain MR Images From (i) Axial Plane (ii) Sagittal Plane (iii) Coronal Plane

Fig.2. Brain MR Images From (i) Axial Plane (ii) Coronal Plane (iii) Sagittal Plane

Different methods and technologies are used for detection and segmentation of cerebrum brain tumor. These technologies are SOM Clustering, K-mean clustering, Fuzzy C-mean.
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It has been observed that finding of brain tumor from MRI images is possible by a variety of methods. The issue of edge discovery is the standout amongst the most appealing issue for the picture preparing because of this current its different applications. Treat boundary identification is the standout amongst the most helpful element in picture division. In this sweet boundary location is utilized for removal of boundaries. F-change is an insightful technique to deal with dubious data. This is helpful for identification of tumor limits. It is exceptionally simple strategy for identification is a promising and productive technique for upcoming and boundary removal advance.

II. BASIC METHODOLOGY

First MRI images of cerebrum brain are taken for processing. The methodology is shown in fig 3. for cerebrum brain tumor detection and segmentation.

**Image Acquisition:** Obtaining of pictures is the fundamental procedure in a picture preparing framework. At the point when the pictures are obtained appropriately then just those pictures can be conceded for pre-handling, division and so on. In the event that pictures are not obtained appropriately, at that point pictures won’t be prepared legitimately. Henceforth acquisitions of pictures are imperative procedure in the picture preparing. First thought about that the fMRI examine pictures of a given patient are either shading, Gray-scale or power pictures in this are shown with a default size of 220×220. On the off chance that it is shading picture, a Gray-scale changed over picture is characterized by utilizing an extensive framework whose passages are numerical qualities somewhere in the range of 0 and 255, where 0 relates to dark and 255 white for example. At that point the mind tumor recognition of a given patient comprises of two principles arranges to be specific, picture division and edge identification.

**Pre-processing stage:** Inconsistencies are the most widely recognized type of clamors that are influenced the nature of the MRI pictures. These things may harm the value of the MR picture subsequently it doesn’t create precise data for progressive picture division and edge identification. So as to maintain a strategic distance from these sorts of value issues, tasks should be performed to evacuate or diminish debasements endured in its securing.

The preprocessing step improves the standard of the cerebrum tumor MR pictures and makes these pictures appropriate for future handling by clinical specialists or imaging modalities. It additionally helps in improving parameters of MR pictures. The parameters incorporates improvement in flag to-clamor proportion, upgrade in visual appearance of MR pictures, the evacuation of superfluous commotion and foundation of undesired parts, smoothing locales of internal part, keeping up applicable edges.

In the first step Pre-processing, noise will be removed by means of different spatial linear or nonlinear filters (Median filter). After noise elimination the next step is text removal by some morphological operations. RGB to dim change and reshaping additionally happens here. It incorporates median filter for clamor migration. The conceivable outcomes of landing of commotion in present day MRI check are extremely less. It might touch base because of warm Effect. Pre-processing steps are shown in figure 4:

**Fig.3. Methodology for cerebrum brain tumor detection and segmentation**

**Fig.4. Preprocessing steps**
Noise Removal and Median Filtering: Middle channel is a non-straight sifting method utilized for clamor expulsion. Middle sifting is utilized to expel salt and pepper clamor from the changed over dim level picture. It changes the estimation of the middle pixel with the middle of the force esteems in the area of that pixel. Middle channels are especially compelling within the sight of drive clamor. Drive commotion is likewise called as salt and pepper clamor on account of its look as white and dark specks secured on picture.

The median filter is use for eliminate salt and pepper commotion from fMRI pictures is shown in figure 5:

![Salt and Pepper Noise](image1.png) ![Median Filter](image2.png)

The middle channel is a successful strategy that can, to some degree, recognize out-of-extent disconnected commotion from authentic picture highlights, for example, boundaries and outline. Particularly the median filter changes a pixel by the median, instead of the common of all pixels in a neighborhood \( \omega \)

\[
Y[m, n] = \text{median} \{x[i, j], (i, j) \in \omega\}
\]

Where \( \omega \) represents a neighborhood defined by the user, centered on location \((m, n)\) in the image.

Various De-noising Filters: The author considered the dissimilar filtering techniques for digital medical image processing that are shown in table 1:

| Various Filters            | Working Rule                      | Advantages                              | Disadvantages                        |
|---------------------------|-----------------------------------|-----------------------------------------|--------------------------------------|
| Mean Filter               | In view of basic estimation of pixels. | Lessens Gaussian Clamor, Answer time is expedient. | Results in hazy confines and limits. |
| Median Filter             | In view of middle estimation of pixels. | Gainful for lessening Salt and Pepper tumult, Spot clamor. Limits and edges are saved. | Troublesome and tedious when contrasted with mean channel. |
| Weiner Filter             | Based on Inverse filtering in Frequency domain | Productive for removing blurring effects from images | Because of working in Recurrence space, its speed is moderate doesn't give fine result for spot clamor. |
| Hybrid Filter             | Grouping of                         | Expels Spot                             | Complex and tedious                  |

Image Smoothing: It is the activity of disentangling a picture while protecting vital data. The objective is to decrease commotion or pointless subtleties without presenting an excess of bending in order to disentangle ensuing examination.

Image Enhancement: To improve contrast the enhancement method is needed. This approach can be used on whole image or part of an image. Differentiation between the cerebrum and tumor locale might be available on a MRI yet not obviously unmistakable through eyes. Along these lines to improve differentiate between the typical cerebrum and tumor district, a high pass channel is connected to the digitized and smooths the MRI which results in better and upgraded picture with plainly unmistakable complexity.

Image Segmentation: The division is mainly imperative stage for breaking down picture legitimately since it influences the precision of the consequent advances. Be that as it may, appropriate division is troublesome on account of the extraordinary verities of the sore shape, size, and hue alongside various skin types and surfaces. Also, a few sores have sporadic limits and at times there is soft change connecting the injury and the skin. To address this issue, a few calculations have been considered. These issues comprehensively named thresholding, area based or edge-based regulated and unsupervised characterization strategies are:

- Water shed segmentation
- Threshold segmentation
- Gradient Vector Flow (GVF)
- K-mean Clustering
- Fuzzy C-means Clustering

Morphological Operations: Morphology manages investigation of shapes and limit territory extraction from mind tumor pictures.
Morphological task is reworking the request of pixel esteems. It works on organizing component and information pictures. Organizing components are properties that test highlights of intrigue. The fundamental activities utilized here are enlargement and disintegration. Enlargement task adds the pixels to limit area, while disintegration expels the pixels from the limit locale of the items. These activities were done dependent on the organizing components. Enlargement picks most noteworthy incentive by contrasting all pixel esteems in neighborhood of information picture depicted by organizing component, while disintegration picks the least incentive by looking at all the pixel esteems in the area of the information picture. Fundamentally the erosion and dilation method are applied for morphological operation.

(a) Erosion: Erosion shrinks or thins substance in a binary image. The way and degree of decrease is restricted by a structuring element. Erosion is a process of translating the structuring element throughout the domain of image. Let A is a set of original image and B is a set of structuring element. Mathematically erosion can be represented as

$$A \odot B = \{ z \mid (B) z \subseteq A \}$$

This condition shows that disintegration A by B is the arrangement of all focuses z with the end goal that B, interpreted by z, is enclosed in A. One of the least difficult employments of disintegration is for disposing of superfluous subtleties from parallel picture. Disintegration is a morphological separating task in which picture subtleties littler than organizing component are sifted from unique picture. The organizing component cover just "1" esteemed pixels of the info. The yield picture has an estimation of "1" at every area of the cause of the organizing component.

(b) Dilation: Widening is a task that develops or thickens questions in paired picture. The way and degree of contracting is constrained by an organizing component. Expansion is a procedure of interpreting the organizing component all through the area of picture. Let A is a set of original picture and B is a set of structuring element. Mathematically dilation can be represented as

$$A \oplus B = \{ z \mid [(B) z \cap A] \subseteq A \}$$

This condition shows that expansion A by B is the arrangement of each uprooting, z, to such an extent that B^A and A cover by somewhere around one component. It is forms that interpret the source of the organizing component all through the space of the picture and check where it covers with "1" esteemed pixels. Translation of structuring element in dilation is similar to that of convolution. One of the simplest applications of dilation is for bridging gaps.

(c) Opening and Closing: In image processing applications, dilation and erosion are used in different combinations. The most common combination are opening and closing. Opening and shutting contains arrangement of enlargements or potentially disintegrations utilizing same or diverse organizing component. Morphological opening evacuate totally districts of an item that can’t contain the organizing component, smooth article forms, break slender associations and expel slim distensions. The Opening of set A by structuring element B is the erosion of A by B, followed by dilation of the result. Morphological closing tends to smooth the contours of object. It generally joint narrow break, fills gap smaller than structuring element.

### III. RELATED WORKS

In [1], automatic brain tumor detection is proposed by using Convolutional Neural Networks (CNN) classification. The deeper architecture design is performed by using small kernels. The weight of the neuron is given as small. Experiment results using CNN shows of 97.5% accuracy with low complication in comparison with other state of arts method.

In [2] the Fuzzy C- Means (FCM) segmentation is applied to divide the cancer and non-cancer region of brain. Also wavelet feature are extracted by using multilevel Discrete Wavelet Transform (DWT). Finally, Deep Neural Network (DNN) is incorporated for cerebrum brain tumor classification with high accuracy. This technique is compared with KNN, Linear Discriminant Analysis (LDA) and Sequential Minimal Optimization (SMO) classification methods. An accuracy rate of 96.97% in the analysis of DNN based brain tumor classification but the complexity is very high and performance is very poor.

In [3] Veeramuthu et. al. characterized as the demonstration of looking at pictures to identify the items. To perceive and treat the brain tumor division can be used to discover and points of confinement of article in pictures. Otsu's thresholding method is used to divide the region of tumor in image, remove features and orchestrated it to keep away from the veritable periods of threatening development.

J. Mehena et. al. characterized an enhancement to the watershed for the removal of mind tumor dependent on partition and morphological administrator. The cancer might be considerable, pre-harmful or threatening and needs medicinal help for further grouping [5].

Presently cerebrum mind tumor is amongst the most risks ailments for recognition ought to be quick and precise. It very well may be accomplished via mechanized tumor discovery strategies on therapeutic images and automated tumor identification procedures are MRI system. It is able to define the tumor expansion area and the edges detection. In the comparison with other approaches, this method is able to provide more accurate with clarity. The advantage of computerized tumor detection technique is [7] exclusion of tumor if needed.

MRI images may have extremely irregular and asymmetrical boundaries in tumor tissues. Deformable modes and region base methods are used for medical image segmentation. After the major finding with the help of MRI images like undefined location of tumor in brain, unobserved borders or data loss at borders and a silent edge then elimination of brain tumor is possible by medical surgery.

In [9] Easha et.al. defines thresholding can’t distinguish cancer
explicitly. The other procedure locale increasing is utilized
which provide seed. Direct methodology to the section ROI
region recognizes cancer effectively. In [10] Komal Sharma
et. al characterized mind tumor is hazardous utilization of
medicinal picture handling in this low quality pictures are
disposed of which have low splendor and commotion. This
concentration to beat the weaknesses in before cerebrum
tumor recognizing strategies and to give better arrangement.

Alan Jose and S. Ravi et.al it gave basic calculation to
location of variety or state of cancer in cerebrum MRI
pictures. The mind tumor identification is prepared
throughout the k-implies grouping and fluently c-implies
calculation to distinguish the precise area and compute the
measure of cancer by figuring the no. of white pixels ideally
[11].

The different approaches with neural networks have been
discovered and used a sizzling research area that was used in
different medical area like a cardiology, radiology, oncology
etc. Highly complex problems can be solved with the
combination of neurons into different layers that makes
artificial neural networks. In the medical image processing
the neural networks like ANNs [13] are used to map an
input into a desired output.

Watershed method is a technique to find cerebrum brain
tumor with actual result by means of accuracy. It is
combined with edge detection operation. Watershed
procedure has capability to provide color cerebrum brain
MRI images. In this method RGB picture changed into HSV
shading picture to convert the picture in to 3 isolated locales
which are known as hue, saturation and intensity. The
vigilant edge identifier is associated to yield a picture for
reconstructed procedure of edge detection that happens in
this procedure. Finally joining [14] the three pictures, we
got resultant cerebrum tumor fragmented picture. This
algorithm is functional over more than 20 cerebrum tumor
MRI images for exceptional outcome.

As per WHO a number of tumors can be categorized on the
basis of central nervous system (CNS). These tumors are
glioma, papillary, glioneuronal tumor etc. The histological
variants are also available that are dissimilar edge
distribution, location, symptoms [16] and the behaviors or
clinical.

Table 2: A Comparative Study for different Cerebrum
Tumor division procedures.

| Segmentation Methods       | Merits                                      | Demerits                                      |
|----------------------------|---------------------------------------------|-----------------------------------------------|
| Region growing Based[3]    | It is finest as it effectively divide        | It is Computationally expensive in states of |
|                            | the districts that have the                | time and memory, receptive to clamor.         |
|                            | comparative properties and give the        |                                               |
|                            | genuine picture which have the             |                                               |
|                            | reasonable limits with great division       |                                               |
|                            | results.                                   |                                               |
| Self-Organizing Map [10]   | It is very uncomplicated, trouble-free to   |                                               |
|                            | recognize and effort incredibly It         | Time consuming                                |
|                            | arranges the information great and         |                                               |
|                            | assesses their very own characteristics,   |                                               |
|                            | with the goal that we can appraise         |                                               |
|                            | fine guide and extreme similarity.         |                                               |
V. RESULT AND DISCUSSION

The author chose MATLAB image processing tool to implement the algorithm because of availability of build functions. The steps of the algorithm are read the gray scale MRI image as in fig 6 and translate it to the binary image as in fig 7 and use median filter to remove noise. The artifacts like text removal will also finished by morphological operations. Furthermore RGB to dim change and reshaping occurs at this time. It integrates median filter for clamor evacuation. The conceivable results of landing of commotion in present day MRI check is not significant. It might touch base because of warm Effect. A series of filters are used to eliminate the image noise. Gaussian Linear filters are used to eliminate image noise. Average Filters are used to eliminate salt and pepper noise from the image. In this filter pixel’s value is replaced with its locality values. Median filter is used to eliminate salt and pepper noise and weighted average filter is the variation of this filter and can be applied straightforwardly and provide significantly improved results. In the median filter value of pixel is determined by the median of the adjoining pixels. This filter is less sensitive than the outliers.

Fig.6: Input MRI image

The tumor portion is detected but there are still holes in the interior of the tumor cell. To fill these holes we use the imfill function. Some morphological operation dilation, erosion is apply on the image. The cell of concern has been segmented effectively. Instead of this object that has been found some other objects that are connected to the border of the image can be removed using the imclear border function. The effect on the images shown in fig 8.

Fig.7: Binary Image

Fig.8: Binary image with filled holes

Fig.9: Segmented image

The outline is created by the bwperim function. Then use a already build function bwarea to compute the segmented tumor region as in fig 9.

V. CONCLUSION

To precisely diagnose the cerebrum brain tumor from the MRI images, an appropriate segmentation and detection method is required on the basis of area of tumor cell calculation. The MRI images are not clear many times and require additional diagnosis. The proposed technique involves preprocessing, enhancement, filtration followed by morphological segmentation. The comparative analysis using GLCM features verifies the efficacy of the proposed technique. The technique may be highly helpful for medical practitioners for accurate and early detection of malignant breast cancers.

Global Thresholding, Genetic Algorithm, and Watershed etc are available for removal and detection of brain tumor. For automatic detection to achieve more accuracy and more efficient methods are deep CNN (Deep Convolutional neural network). Deep CNN is one of the profound learning strategies, which contains sequence of feed forward layers and gives better accuracy and reduce the computation time.

REFERENCES

1. Seetha J, Raja S. S., “Brain Tumor Classification Using Convolutional Neural Networks,” Biomedical and Pharmacology Journal. Vol. 11(3). p. 1457-462’ (2018).
2. Heba Mohsen et al, “Classification using Deep Learning Neural Networks for Brain Tumors”, Future Computing and Informatics, pp 1-4 (2017).
3. Veeramuthu, S. Meenakshi, and V. P. Darsini, “Brain image classification using learning machine approach and brain structure analysis,” Procedia Comput. Sci., vol. 50, pp. 388–394, 2015.
4. Khan and Z. Amari, “Soft Computing based Medical Image Mining: A Survey,” Int. J. Comput. Trends Technol., vol. 27, no. 2, pp. 76–79, 2015.
5. J. Mehena and M. C. Adhikary, “Brain Tumor Segmentation and Extraction of MR Images Based on Improved Watershed Transform,” vol. 17, no. 1, pp. 1–5, 2015.
6. Brundha and N. K. M., “MR IMAGE SEGMENTATION OF BRAIN TO DETECT BRAIN TUMOR AND ITS AREA CALCULATION USING K-MEANS CLUSTERING AND FUZZY C-MEANS ALGORITHM,” vol. 2, no. 9, pp. 1781–1785, 2015.

7. S. Chrutha and M. J. Jayashree, “An efficient brain tumor detection by integrating modified texture based region growing and cellular automata edge detection,” in Control Instrumentation, Communication and Computational Technology (ICCICCT), 2014 International Conference On, 2014, pp. 1193-1199.

8. R. Preetha and G. R. Suresh, “Performance Analysis of Fuzzy C Means Algorithm in Automated Detection of Brain Tumor,” in Computing and Communication Technologies (WCCCT), 2014 World Congress on, 2014, pp. 30-33.

9. Easha Noureen and 2Dr. Md. Kamrul Hassan, “Brain Tumor Detection Using Histogram Thresholding to Get the Threshold point/α/β,” IOSR J. Electr. Electron. Eng., vol. 9, no. 5, pp. 14–19, 2014.

10. K. Sharma, A. Kaur, and S. Gujral, “A review on various brain tumor detection techniques in brain MRI images,” vol. 4, no. 5, pp. 6–12, 2014.

11. Jose, S. Ravi, and M. Sambath, “Brain Tumor Segmentation Using K-Means And Fuzzy C-Means Algorithms And Its Area Calculation,” Int. J. Innoy, Res. Comput. Commun. Eng., vol. 2, no. 3, pp. 3496-3501, 2014.

12. J. Patel and K. Doshi, “A Study of Segmentation Methods for Detection of Tumor in Brain MRI,” Adv. Electron. Electr. Eng., vol. 4, no. 3, pp. 279–284, 2014.

13. A. Abdullah, et al., “Implementation of an improved cellular neural network algorithm for brain tumor detection.” in Biomedical Engineering(Isobel), 2012 International Conference on, 2012, pp. 611-615.

14. N. Louis, et al., “The 2007 WHO classification of tumor of central nervous system,” Acta neuro pathological, vol114, pp 97-109, 2007.

15. P. Kleihues, et al., “The new WHO classification of brain tumor,” brain pathology, vol3, pp. 255-268, 1993.

16. https://www.researchgate.net/profile/Arjun_Nichal2/publication/317547838/figure/fg1/AS:505110660907008@1497439174818/Basic-Block-diagram-of-brain-tumor-detection-and-segmentation.png.