

# Review of Brain Tumor Detection Concept using MRI Images

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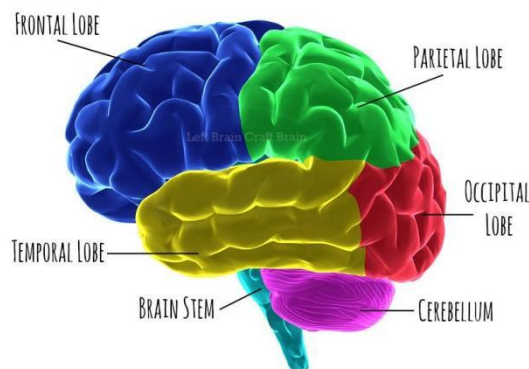
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**Abstract:** *Magnetic resonance imaging (MRI) processing is one of the components of image processing in the medical profession, which has recently experienced the largest growth. The growth Detection is frequently the first stage. This study explains the thresholding method for brain tumor identification. The suggested method can effectively be used to recognise and isolate brain tumors from MRI images collected from patient databases. It functions as a useful tool for doctors who practice in this area. Grayscale image, brain tumor, MRI, edge detection (using the sobel operator), filtering, thresholding, and shrinking operation on image are the keywords. Given the complex structure of the brain, detecting brain tumors with MRI images is a difficult undertaking. It is possible to partition the brain tumor and prepare X-beam images.*

**Keywords:** Magnetic resonance imaging

## I. INTRODUCTION

The brain is one of the most crucial parts of the human body since it regulates the operation of all other organs and aids in decision-making[10]. It is basically the central nervous system's command post and is in charge of carrying out the body's regular voluntary and involuntary functions. [1] The tumor is an uncontrolled proliferation of undesirable tissue that has formed a fibrous mesh inside of our brain. A brain tumor is identified in roughly 3,540 youngsters this year at the age of 15 [9]. To effectively prevent and treat the condition, it is crucial to have a thorough grasp of brain tumors and their stages. This study uses image planning tools to reverse the obvious proof of a brain tumor. The work is broken up into five sections, each of which discusses a different component of brain tumor diagnosis by image processing. Presenting Sections II and III background and reason for employing image processing to find brain tumors. The available related work and research are discussed in Section IV. Section V provides a technique description and acknowledges the research. However, because MRI uses magnetic waves, people with pacemakers and metal implants should avoid getting one. Once the brain has been scanned, it is crucial to precisely identify the tumor, its size, and its location. The neurosurgeon needs all of this information to complete his diagnosis. Computational image processing can be useful in this situation.

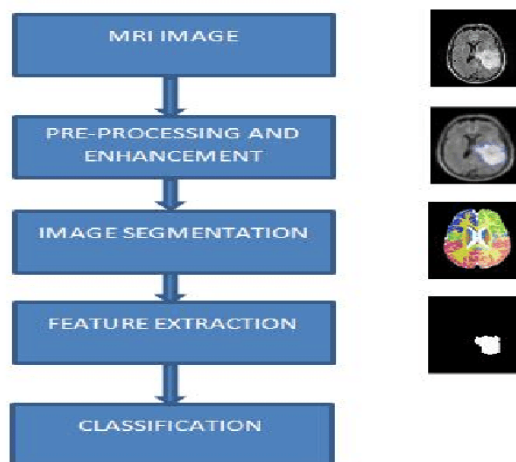


We can precisely identify the tumor by utilizing several segmentation approaches and feature extraction methods. In this study, the classification of normal and malignant brains using ANN and CNN. [10] A vast number of interconnections

and networking enable ANNs (Artificial Neural Networks), which mimic the functioning of the nervous system in the human brain, to train using only basic processing units on training data and store learned information. It has numerous layers of neurons that are interconnected. The neural network can learn by [3] applying data sets to the learning process. There will be one input and output layer however there may be any number of hidden layers.

## II. METHODOLOGY

Thresholding, the use of morphological operations, and the extraction of the tumor region for additional analysis are the primary pillars of the suggested study. The application of the constant "T" over an image is the thresholding technique. Global thresholding is the act of applying the threshold constant T across the entire image, while variable thresholding is the process of altering the value of T across an image. The MRI scanning process yields the brain image. MATLAB R2014a has been used to implement the experiment.



### 2.1 Image Pre-Processing:

The act of processing an image is challenging. Remove any unwanted elements from any image before it can be processed. There may be unwanted artifacts there. The image can only be effectively processed after that. Two basic phases are involved in processing a medical image. The first step is image pre-processing. To prepare the image for the next stage, this requires executing operations like noise reduction and filtering. Segmentation and morphological operations are carried out in the second step. They define the tumor's size and placement. A. Image Pre-Processing: To ensure that the results of our image processing are accurate, it is critical that the image be free of any unwanted data and in the proper format. Pre-processing is the name given to this preparation. Pre-processing entails stages like grayscale conversion, noise reduction, noise removal, image enhancement, and, in the case of medical imaging, measures like removing the skull from an MRI. The image is filtered to remove further noise after it has been transformed to a grayscale image. There are two different kinds of filters: those that let high-end frequencies pass and those that let low-end frequencies pass. An image can be flattened or sharpened with a filter. The image's finer details are lost when it is flattened by a filter, which also blurs the image's noise and leaves behind a smooth image. If the image is sharpened, the filter brings out the finer features, but doing so increases the amount of noise in the picture. Prior to further processing, this noise should be removed because it may interfere with the detection program's accuracy. A grayscale image, on the other hand, is made up entirely of different tones of gray. This means that each pixel does not display any color and instead represents the intensity value present at that pixel. A grayscale image, in contrast to a black and white image, has numerous shades, with white being the lightest shade and black being the darkest.

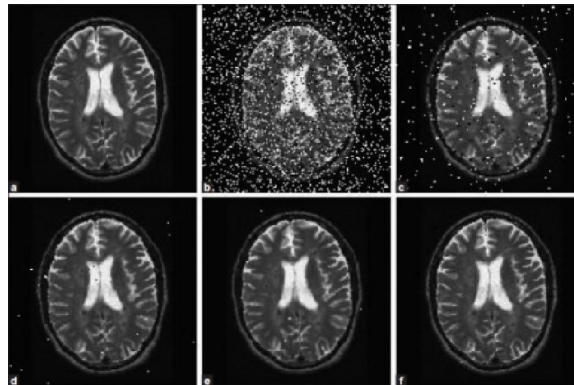
#### Filter: Mean

This is the common disposal technique that is most well-known. This separating method is 'non-straight'. This is used to remove the "Salt and Pepper" structure from the grayscale image [10]. Middle channel relies on a typical pixel estimate. The upsides of the middle channel are useful in minimizing Salt and Pepper disturbance and Speckle noise.

The edges and limitations are also preserved. When compared to a mean channel, the main obstacles are the multifarious character and time usage.

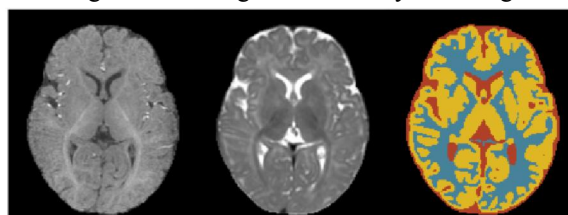
**Filtering: Median Filter**

one of the methods for noise cancellation most frequently used nowadays. The basis for itsIts common use is because it keeps the image's edges intact. According to what the name implies, the median of each entry's adjacent entries is used in its place. This filter is particularly good for reducing salt and pepper noise and poisson's noise. This filter operates by sweeping the entire signal in a pattern. The output intensity is determined by the median pixel intensity in the pattern. A median filter was chosen by Drs. M. Karnan, A. Lakshmi, and A. S. Bhalchandra to eliminate noise from their experiment. A pseudo color translation technique was put out by Ming-Ni Wu[4] to be used on the grayscale image. A pseudocolor image, also known as a false color image, differs from a conventional image in that coloured items are detected differently here than they would be in the genuine image. By giving each intensity value a color based on a function, a grayscale image is transformed into a pseudo-color image. Thermal imaging is an example of a pseudo-color image. When there is just one data channel available, it should be used.



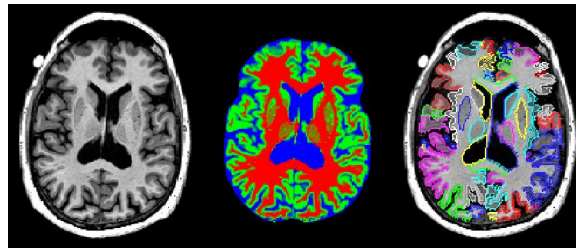
**Segmentation:**

Based on the similarity of the regions, a process called "mage segmentation" separates an image into relevant regions. Many applications, including image analysis, object detection and representation, and region of interest visualization, use image segmentation [8]. It facilitates image analysis for specific purposes. Since most medical images have poor contrasts and may be affected by noise or diffuse borders, image segmentation is crucial in the field of medicine [9]. Noise removal is a stage in the preprocessing of these images [10]. Two essential characteristics of picture intensity levels are used by image segmentation algorithms: image discontinuity and image similarity [11].



Therefore, segmentation is carried out in two steps. The first modifies a pixel's intensity, such as its brightness at edges and corners, in an image. Based on dividing an image into regions, the second one. There are a variety of segmentation techniques based on these that are widely utilized in a variety of applications. A brand-new FCM-based technique for spatially coherent and noise-resistant picture segmentation was presented by Ivana Despotovi. It offers two significant contributions. First, to account for the effect of noise, local image characteristics' spatial information is incorporated into the similarity metric and the membership function. Next, an anisotropic neighborhood based on phase congruency features is implemented to enable more precise segmentation without image smoothing. The segmentation results show that this method effectively maintains the homogeneity of the regions for both artificial and real images. It is the approach of scattering a picture into smaller fragments. It arranges the pixels in different ways inside the same image.

gives each pixel in a photo a tag, and pixels with the same mark share certain highlights [11]. It is also simpler to deconstruct and understand important data structures in a complex image when the image is fragmented. Images are frequently segmented using a variety of clustering techniques. A group of pixels with some shared traits is referred to as a cluster. It entails categorizing items according to how similar they are to one another. Clustering methods can be further broken down into two categories. According to hard type clustering approaches, an object can only be a member of one cluster. This leads to very clear segmentation. Unfortunately, this technique becomes quite challenging if the image has poor resolution and contrast. The K-Means Clustering Algorithm is an illustration of a hard clustering technique. With each object belonging to the cluster with the closest mean, the K-Means method seeks to divide the image into n partitions. Each point in this case can only belong to one cluster.



### Feature Extraction

It is the tactic of breaking up a picture into small pieces. It arranges the pixels in different ways inside the same image. gives each pixel in a photo a tag, and pixels with the same mark share some highlights [12]. It is also simpler to deconstruct and understand important data structures in a complex image when the image is fragmented.

### III. CONCLUSION

A cerebrum tumor is defined as an unpredictable growth of brain tissues that affect real psychological boundaries. The main goal of clinical picture analysis is to identify precise and important information from photos with the least amount of mistake feasible. Due to the complexity of the brain, detecting clear signs of a brain tumor through MRI images is a difficult undertaking. There are numerous image division approaches that can be used to segment these malignancies. Pre-treatment, picture division, feature extraction, and picture request are the four distinct steps in the process of identifying brain tumors in MRI images.

The many methods used to identify brain cancers from MRI images are covered in this paper's examination of medical image processing techniques. The various techniques currently employed in medical image processing were first thoroughly researched. Studying the available research was required for this. This document, which lists the several methods used, was written using the study mentioned above. There is also a brief explanation of each procedure. Segmentation is also the most important and advantageous of all the many processes involved in the process of finding tumors.

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