Image Processing Techniques for Brain Tumor Identification

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Abstract. At present, treatment of patients requires clinical pictures examination and is becoming a critical field. It fuses a wide scope of imaging procedures Some of them are Computed Tomography channels (CT examines), Magnetic Resonance Imaging (MRIs) and X-columns and so forth. These progressions help us to perceive even the slightest imperfections in the human body. Flighty development of tissues in the mind which sway genuine cerebrum limits is considered as a cerebrum tumour. The standard objective of clinical picture dealing with it is to see precise and vital data utilizing pictures with the base mess up conceivable. X-pillar is overwhelmingly used to get photographs of the human body and risky tissues as a result of its noteworthy norms and better-quality pictures separating other imaging headways. Brain tumor conspicuous bits of proof through MRI pictures is an irksome task because of the multifaceted design of the cerebrum. X-beam pictures can be readied and the psyche tumor can be partitioned. These tumors can be separated using diverse picture division techniques. The path toward recognizing cerebrum tumors through MRI pictures can be masterminded into four extraordinary territories; pre-planning, picture division, incorporate extraction and picture gathering.

Keywords: Brain Tumor, Image Processing, Feature Extraction, Classification, Convolutional Neural Network.

1. Introduction
Mind tumors sway the people brutally, on account of the uneven growth of cells inside the cerebrum. It can agitate valid cerebrum work and be unsafe. Two kinds of cerebrum tumors have been seen as Benign tumors and Malignant tumors. Benevolent tumors are less subverting than threatening tumors as they are quick making and perilous while the past tumors are moderate getting and less risky.
Clinical imaging system is used to make visual outline of inside the human body for clinical purposes and non-noisy possibilities can be dissected by this movement. The various types of clinical imaging degrees of progress reliant on non-conspicuous methodology like: CT channel, SPECT, MRI. Ultrasound, PET and X-segment \cite{1}. When showed up unmistakably practically identical to other clinical imaging frameworks, Magnetic Resonance Imaging (MRI) is essentially used and it gives more clear lightness photos of the cerebrum and hurting tissues. In this way, mind tumour perceiving check should be conceivable through MRI pictures. This paper turns around the conspicuous proof of cerebrum tumour using picture planning systems. The paper is divided into V sections and describes various aspects of identification of brain tumour using image processing. Section II and III present motivation and background for identifying brain tumours using image processing techniques. Section IV discusses the related work and research available. Section V describes the methodology used and also gives an acknowledgement of the research.

2. Motivation
Automated gathering and area of tumors in different clinical pictures is moved by the need of high accuracy while dealing with a human life. Moreover, the PC help is mentioned in clinical foundations in light of the way that it could improve the delayed consequences of individuals in such a space where the sham sceptical cases must be at a particularly low rate. It has been exhibited that twofold scrutinizing of clinical pictures could provoke better tumor ID. Regardless, cost recommended in twofold examining is amazingly high, that is the explanation adequate programming to help individuals in clinical establishments is of unfathomable interest nowadays. Conventional procedures for checking and diagnosing the contaminations rely upon recognizing the closeness of explicit features.
by a human onlooker. Due to gigantic number of patients in concentrated thought units and the necessity for constant impression of such conditions, a couple of techniques for automated decisive structures have been made starting late to try to deal with this issue. Such methods work by changing the generally emotional demonstrative guidelines into a more objective quantitative segment gathering issue. In this paper the automated request of cerebrum MR pictures by using some previous data like pixel power and some anatomical features is used.

3. Background
Brain Tumor is depicted as surprising development of tissues in the cerebrum. These days the lack of ability of tumors is developing rapid. In 2016, an ordinary 23,800 grown-ups in the US was related to the unsafe tumors of cerebrum comparatively as spinal code [2]. Assessment of cerebrum tumors is to some degree questionable as the moved shape, size, tumor region and the closeness and proximity of tumor at the head of the need list. It's difficult to perceive cerebrum tumors in early phase considering the way that the particular estimation of tumor can't be found. By the by, when the cerebrum tumor is perceived at doubtlessly the early phase, the best medications ought to be possible and it might be reparable. At present, visual portrayal of inside the body is prepared utilizing clinical imaging technique for clinical appraisal and clinical explores. X-pillar is the best and comprehensively utilized procedure for mind tumor region. Current finding techniques are performed utilizing the normal systems dependent on human experience and this develops the chance of bogus zone while seeing cerebrum tumors. Present devices and frameworks to eviscerate tumors and their lead have gotten powerfully otherworldly. Picture preparing method can be utilized to see cerebrum tumors. Picture preparing techniques changes over pictures into front line and do framework on them, so as to give signs of progress and overhauled pictures [3]. This evaluation will think how to see cerebrum tumors utilizing picture dealing with methodologies.

4. Related work
In recent years, image processing has applied to process pictures in clinical stream, in coordinating cell identification. S. Mokhle in 2012 [3] presented a few distinguishing proof advances, including fragmenting pictures to extricate the item from the foundation through the edge. This element was presented with the 'Gabor channel' so as to accomplish more arrangement into malignant growth cells. H. G. Zadeh in 2013 [4] proposed further advances, which is picture extraction and division of pictures for diagnosing malignancy cells. The Gaussian smoothing idea was presented as a separating reason, past to applying the 'Quick Fourier Transform' (FFT).

AI for tumor discovery: 'NN', 'Fluffy C-signify' calculations was presented for the recognizable proof of tumorous cells [6]. This takes lower computational time however the precision additionally lower. X. Chen presents quality checking innovation in 2014 [5]. Be that as it may, this innovation is fitting just for the perplexing development of quality choice. From the previously mentioned methods and utilizing of advances, in this examination paper we centre around the recognizable proof of mind tumor utilizing picture handling procedures.

5. Methodology
By using the following steps, we can detect Brain tumors using Image Processing techniques.

![Image processing steps diagram](image_url)
5.1 Pre-Processing of Image
It is a bit hard to process a picture. Before any picture is handled, it is exceptionally noteworthy to evacuate pointless things it might hold. Subsequent to expelling superfluous curios, the picture can be prepared effectively. The underlying advancement of picture handling is Image Pre-Processing [9]. Pre-Processing includes forms such as transformation to greyscale picture, commotion evacuation and picture reproduction. Change to greyscale picture is the most widely recognized pre-preparing practice [8]. After the picture is changed over to greyscale, at that point overabundance commotion is expelled utilizing diverse separating techniques.

![Figure 2. Image enhancement.](image1)

5.1.1 Filter: Mean
This the most widely recognized method which is utilized for commotion disposal. This is a 'non-straight' separating strategy. This is utilized to wipe out 'Salt and Pepper clamor' structure the greyscale picture [10]. Middle channel depends on normal estimation of pixels. The upsides of middle channel are effective in decreasing Salt and Pepper commotion and Speckle clamor. Additionally, the edges and limits are saved. The fundamental hindrances are multifaceted nature and time utilization when contrasted with mean channel.

![Figure 3. Application of Median filter](image2)

5.1.2 Filter: Median
This channel is additionally a de-noising channel that depends on normal estimation of pixels. Major interest of mean channel is it lessens Gaussian clamor and the reaction time is quick. Primary burden is its contorted limits and edges.

5.1.3 Filter: Wiener
The Wiener Filter is additionally a de-noising channel that depends on the converse sifting in the recurrence space. Productive to dispose of pictures as obscure is the primary favorable position of the Wiener Filter. As it is working in the recurrence area, it has less speed and isn't reasonable for Speckle commotion.

5.1.4 Filter: Hybrid
The Hybrid channel comprises both Median channel and Wiener channel. It can dispense with Speckle commotion, Impulse clamor and obscuring impacts from pictures. The multifaceted nature and time utilization is the fundamental impediment of the Hybrid channel.
5.1.5 Filter: Modified Hybrid, Median
This filter is likewise a de-noising channel which contains both Median and Mean channel. This is productive to wipe out Speckle commotion, Salt and Pepper clamor and the Gaussian clamor. However, the fundamental disservice of this channel has high time utilization contrasted with the basic Median channel.

5.1.6 Filter: Morphology, Based De-noising
This channel depends on Morphological activities of opening and shutting. Delivering results better than other de-noising channels and the effectiveness are the primary bit of leeway of this channel.

5.2 Image Segmentation
It is the strategy of dispersing a picture into smaller bits. It makes a few arrangements of pixels inside same picture. Does out a tag to each pixel in a picture and the pixels with the comparable mark share specific highlights [11]. Fragmenting makes it simpler to additionally break down and perceive significant data structure an advanced picture.

5.2.1 Threshold Segmentation
'Segmentation' is the method that has been used to separate a computerized image into many parts that incorporate arrangements of pixels and collection of super pixels. Goals to be cultivated through the procedure of division are rearranging and changing the arrangement of portrayal of a picture such that it will turn out to be increasingly nitty gritty, significant and simple for the procedure of investigation. Putting of articles and limits in pictures, for example, lines, bends could be performed through Image division. All through the method of picture division, each pixel in a picture is doled out a name and the pixels comprise of same mark share certain visual highlights. Every pixel in the district is comparable according to certain highlights or figured properties, for example, shading, force or surface. Bordering locales are especially unique with respect to similar highlights. Threshold strategy is the most straightforward method of picture division. This procedure includes an edge esteem that is accustomed to changing over a dark scale highlighted picture to a paired picture. The significant bit of leeway of this technique is choosing the edge an incentive to be utilized.

5.2.2 Segmentation Based on Morphology
'Morphology' allows to identify the properties of the structure and shape of any element. Paired pictures may contain numerous deformities. Especially, the parallel locales developed by basic thresholding are distorted by surface and commotion. Morphological picture preparing looks to accomplish the objectives of dispensing with these imperfections by representing picture shape and structure. For the most part, this signifies perceiving items or limits inside the picture.

Morphological tasks are intelligent transformations dependent on correlation of neighbouring pixel. Typically, morphological tasks are actualized on parallel pictures under the pixel esteems; 0 or 1. A significant number of the morphological activities focus on twofold pictures.

5.2.3 K-Means, Algorithm
Majority of picture handling procedures use K-Means calculation for picture division. It is extremely valuable for enormous pictures with helpless complexity. In any case, it has been understood that K-Means is vulnerable to determination of tests and foundations of fluffy sets [10], [11].

\[ J = \sum_{j=1}^{K} \sum_{i=1}^{n} \|x^{(i)}-c_j\|^2 \]  \hspace{1cm} (1)

\[ J = \sum_{j=1}^{K} \sum_{i=1}^{n} \|x^{(i)}-c_j\|^2 \]  \hspace{1cm} (1)
5.3 Feature Extraction

It is the strategy of dispersing a picture into minor bits. It makes a few arrangements of pixels inside same picture. Does out a tag to each pixel in a picture and the pixels with the comparable mark share specific highlights [12]. Fragmenting makes it simpler to additionally break down and perceive significant data structure an advanced picture.

5.3.1 Detection of edge

An edge detection happens when there is an abrupt and surprising force adjustment of the picture. At whatever point it is identified a sudden alteration or a change in the power of a specific picture, the related pixel would be treated as an edge pixel. The calculation that has been advanced for the discovery of edge pixel underpins in distinguishing the nature of the edge. Be that as it may, some of the time these edges are not shown in the conclusive outcome. Henceforth the calculations are acclimated to decide the edges [16].

5.3.1.1 Edge detection: 'Prewitt'

The "Prewitt Mask" is one of the unmistakable separation activity. As needs be, approximated subsidiary qualities in both the headings, with the end goal that even and vertical, are determined utilizing two $3 \times 3$ veils [16]. Prewitt veils provide an approximation to both flat subsidiary and the vertical subordinate.

5.3.1.2 Edge detection: 'Robert'

By using the "Roberts edge" location activity, the picture inclination is evaluated by means of unmistakable separation. Likewise, "Robert Mask" is a network and the districts of high spatial recurrence are featured, that are frequently compared to edges in the image [16].

5.3.1.3 Edge detection: 'Sobel'.

The "Sobel Mask" generally function as the "Prewitt veil". It must be taken into account as the Sobel administrator has values; '2' and '-' 2' which are assigned to the focal point of first and the third segments of the flat veil and first and third columns of the vertical cover. Henceforth it gives high edge intensity [16].

5.3.2 Feature Extraction using Histogram of Oriented Gradient

The extraction technique of the "Histogram of Oriented Gradient" (HOG) will take the calculations given below in consideration. In the first place, the pre-arranged cell picture shall be coursed as "$32 \times 32$" pixels. The power of

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**Table 1: Different segmentation methods**

| Methods for Segmentation | Used parameter | Prone towards |
|-------------------------|----------------|---------------|
| Threshold method        | gradient magnitude is applied to find the potential edge pixels [11]. | Difficult to be applied to pictures having low contrast. |
| Region Based method     | Used for separation of regions in an image based on similar properties [13]. | Quality of final result may be impacted by noise. |
| Fuzzy C Means, K Means and Level Set Techniques | Can be applied to pictures with large size and poor contrast. Establishing of fuzzy sets and selecting sample may be bit hard [15]. | |
each pixel is '1' or '0'. By then the result will be added to "Crowd" [17]. Figure. below shows the building of "Crowd" incorporate. By then the image shall be broken into "8 × 8" pixels that is called matrix. Here, the matrix shall be now included into a single square. Once more, each case shall be coursed into 9 canisters which is "3 × 3". Pixel edge is used for the creation of the component in each and every container [17]. As there are a total of 9 features, the vector image will incite "9 × 4" characteristics for each square. In the each of the "32 × 32" pixels, "Crowd" incorporate extraction licenses to make '9 squares in conclusion, it must have "9 × 9 × 4" features in one estimation or "1 × 324".

![Diagram](image.jpg)

**Figure 4.** HOG feature architecture

6. Discussion
The most critical in picture handling is picture division, while diagnosing mind tumor from an advanced picture. Table 2 speaks to the distinctive division strategies with various attributes, for example, exactness and unpredictability.

| Segmentation Method       | Complexity Algorithm | Accuracy % |
|---------------------------|----------------------|------------|
| Seed region growing       | 10                   | 92.5       |
| Threshold segmentation    | 8.22                 | 91         |
| Water Shed                | 5.67                 | 88.5       |
| Fuzzy C-mean              | 5.29                 | 85         |
| Histogram                 | 7.61                 | 81         |

Primary objective of Pre-Processing is the edge conservation of the picture. Among the edge discovery systems, Sobel is the most ideal choice then both the Gaussian and the Median channel. The distinctive edge discovery calculations like 'Prewitt', 'Sobel', and 'Robert' can be used with various attributes and are discussed below in Table 3, for example, calculation time in short order and calculation count of flip-flops utilized.
Table 3. Edge detection method: A performance comparison

| Method Name | Advantages                        | Disadvantages            | Computation in seconds | Number used as flip flops | Number used as logic |
|-------------|-----------------------------------|---------------------------|------------------------|--------------------------|---------------------|
| Sobel       | Simplicity                        | Discontinuity in edges    | 0.3                    | 343                      | 450                 |
| Robert      | Better Noise suppression          | Not accurate results      | 0.2                    | 219                      | 322                 |
| Prewitt     | Mask simpler as compared to Sobel | Discontinuity in edges    | 0.4                    | 339                      | 450                 |

The accompanying figures exhibit the means of cerebrum tumor discovery utilizing picture handling strategies. That is unique MRI picture, grayscale picture, sifting picture utilizing Median channel, fragmenting utilizing edge strategy, morphological activity applied picture lastly analyzed tumor from MRI image segmentation procedures. It simple to execute and broadly utilized nowadays.

![Figure 5. Brain tumor detection Steps](image)

At the point when the difference between frontal area item and foundation object is similarly high, limit procedure functions admirably.
7. Conclusion & Applications

Unpredictable improvement of tissues in the cerebrum which impact authentic psyche limits is considered as a cerebrum tumor. The key target of clinical picture taking care of is to recognize exact and significant information using pictures with the base error possible. Brain tumor unmistakable bits of proof through MRI pictures is an inconvenient task because of the multifaceted idea of the cerebrum. These tumors can be divided using diverse picture division methodologies. The path toward recognizing cerebrum tumors through MRI pictures can be masterminded into four unmistakable sections; pre-taking care of, picture division, feature extraction and picture request.

Middle channel is the most generally utilized separating procedure among different sifting strategies. Less multifaceted nature and the productivity in disposing of 'Salt and Pepper clamor' are the primary points of interest of middle channel.

Unlike Gaussian channel, it is a non-direct channel, Median channel is an edge sparing channel. Moreover, Gaussian channel is a low pass channel from this time forward the edge information will be lost and edges getting removed and clouded. Though, less multifaceted nature and the reasonableness to complete than the Median channel are the major focal points of Gaussian channel. Another piece of space is the Gaussian direct is altogether material in smoothing Gaussian uproar. Thresholding is the best and most clear system among picture.

Using this methodology, early stage brain tumour detection is possible. Tumour detection can be done with increased accuracy at high speed. It provides automation in the field of image processing as well as analysis and improves identification of brain structures in medical science.

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