

Brain Tumor Prediction through Behavior Analysis of Cells Growth Using Machine Learning Techniques

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Abstract-Brain tumor is a very terrible disease. Brain tumor is caused by an increased number of cells. The presence of the skull layer around the brain makes it tough in studying the behavior of growth cells. It also raises the complication for the identification of disease. The initial discovery of a brain tumor is necessary to defend the survival of patients. Frequently, the brain cancer segmentation, and classification through the MRI images technique. Though, the radiologists are not providing actual visualization of brain cells in MRI images due to the irregular growth of cells, which forms of cells are growing rapidly and slow at some stage in brain tumors in the brain. So, automatic strategies are required to evaluate thoughts tumors exactly from MRI images in this research automatic, MRI brain tumors are used for classification, segmentation, and Behavior analysis of cell growth. The problem of visualization of cell growth and behavior analysis of brain cells is solved through MRI images which enhance the detection of cancer. To analyze the behavior of cell growth, which forms of cells are growing rapidly and slow at some stage in brain tumors, and analyze the area of images in which type of cells is affected. Single models are less efficient. We will use ensemble models which would also be helpful for better performance and accuracy.

Keywords: Index Terms, MRI, CNN, SVM, Detection, Cancer.

I INTRODUCTION

The brain, which controls the rest of our organs, is the most essential element of the human body. Brain tumor have been ranked as the world's second leading cause of death [1]. People, on the other hand, have not been aware of them and have not received appropriate therapy in a timely manner. Many applications of computer revelation make human life much easier and safer. Image processing techniques and machine learning approaches is the real of medical pictures aspire to make medical diagnosis easier[2]. The expert's review and evaluation of all MRI pictures in order to determine whether they include brain pathology or not is a time consuming process. Computer Aided Diagnosis (CAD) of checkup photos detects brain pathology earlier than a human consultant, which is quite beneficial to them. Magnetic Resonance Imaging (MRI), X-ray Computed Tomography (CT), and Positron Emission Tomography (PET) scans can all be used to get medical images of the brain[3]. The brain tumor is one of the maximum terrible natures of most cancers and triggered a large number of deaths among children and adults the few

years according to the world health organization (WHO) standard, 700,000 human beings are existence with mind cancer and around 86,000 are diagnosed considering that 2019[4]. Two kinds of brain cancers had been recognized benign tumors and malignant tumors. Benign cancers are much a reduced amount of harmful than malignant cancers as malignant are speedy developing and harmful whereas benign are slow-growing and much less harmful. Sometimes the tumors are placed in sensitive regions of tears[5]. The main intention of medical image processing is to become aware of accurate and significant facts and the use of an image with the minimal errors possible[6].

A. Objectives

- The main goal is to classify normal and abnormal cells through MRI images.
- Also, detect the area of cancer and analyze the speed of cells.
- Analyze the behavior of cell growth, which forms of cells are growing rapidly and slow at some stage in brain tumors.
- We will achieve more and better accuracy through visualization.

B. Problem statement

The brain is the most complex structure in the human physique. A brain tumor is a very dangerous infection. The expanded growth of cells is causing the brain tumor. The presence of the skull layer around the brain makes it tough in studying the behavior of growth cells and also raises the difficulty for the identification of disease. It causes many serious problems and also causes death. The main reason behind the brain tumor is that it cannot be predicted or visualized at an early stage. Because of this, normal and abnormal cell growth cannot be controlled. This research work solved the problem of visualization of cell growth and behavior analysis of brain cells through MRI Images.

C. Research Questions

1. How does the behavior of growth cells become easy to identify using MRI images?
2. How to can classify healthy and unhealthy cells?

3. Why single models do not perform visualization of brain cells more accurately?
4. How to identify the rapid growth of cells during brain tumors?

D. Our contributions

- Classification of tumor and non tumors images using ensemble learning.
- Segmentation of timorous stages.
- To analyze, which forms of cells are growing rapidly and slow at some stages in brain tumors, and analyze the area of images in which type of cells is affected.
- To achieve better accuracy we will use the ensemble learning.

II LITERATURE REVIEW

The mind is a significant structure of a human being’s sensory system. For the analysis of brain growth, MRI is used. Numerous semi-automated and completely automated methods are presented in the literature for brain cancer detection and visualization of mind cells[7]. Deep learning is a developing knowledge in machine learning and has involved important considerations in medical image processing specifically for the visualization of brain cells[8]. Then, at that point, CNN grouping is applied for the segmentation of brain cells. Grayscale level run length matrix (GLRLM) is utilized in getting highlights starting the brain picture, at that point, the SVM method is into a characterization of brain MRI pictures[9].

The size of the phase of cancer is recognized by utilizing data set frameworks strategy [10].It influences the sensory cells, synapses, organs, and films that encompass the mind. Growth can be analyzed utilizing imaging and pathology. Brain growth is imaged with MRI (Magnetic Resonance Imaging), generous cross-segment pictures of the brain[11]. The author has developed a model dependent on Artificial Convolution Neursal Networks that takings these attractive reverberation pictures furthermore, investigate them with numerical recipes and framework tasks. This neural organization gives us the likelihood of how possible the presence of growth in the mind, and had prepared attractive reverberation pictures[12].

III METHODOLOGY

This study presents a visualization of brain cells and behavior analysis through MRI images. Firstly, give input to the MRI images. The second step is pre-processing MRI images. The sub-set of pre-processing is skull stripping, resizing the image, and image filtration. The next step is segmentation on the region of interest, in this step to extract the relevant things in the brain tumor. The system presents a ratio of almost 80% training and 20 % testing. The classifier model is used to achieve a better accuracy model. The last step is to visualize the classified data to find normal cells and abnormal cells. The main goal is to analyze the behavior of cell growth, which forms of cells are growing fast and slow at some stage. The speed of cells connecting the brain tumor types. Through visualization of classified data, the data will appear in graph form and show the accuracy of cells speed we will find normal cells or abnormal cells. And predict the brain tumor. This methodology of brain tumors

through behavior analysis of cell growth is shown in the figure 1. In this manner, we will give input of 32x32 to the model, then images are extracted by convolution, sub sampling, and featured maps are shown accordingly based on 28x28 feature maps having 5x5 convolution, 14x14 feature maps having 2x2 sampling, 10x10 feature image having 5x5 convolution, 14x14 feature maps having 2x2 convolution behaving sub sampling. CNN classifier is applied to the extracted data after detection through different pre-processing techniques. CNN model is used for experimentation. CNN will classify the benign and malignant tumors in the last step as shown in figure 2.

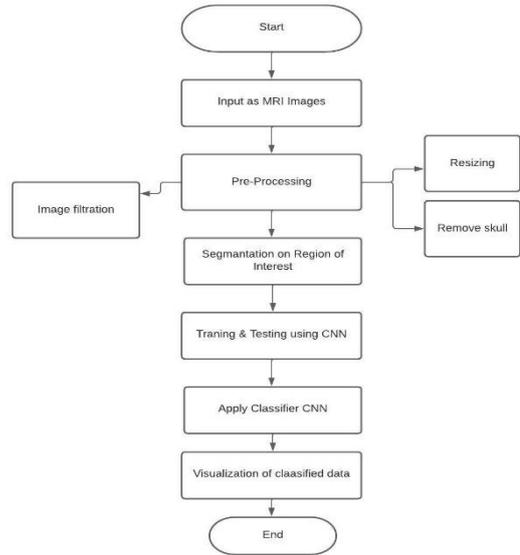


Fig1: Brain tumor prediction Methodology

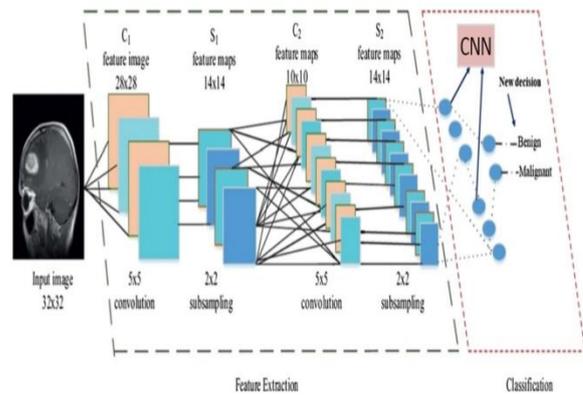


Fig2: Structure of CNN classification the benign and Malignant tumors

IV RESULTS AND DISCUSSIONS

In trendy, mind photo segmentation methods are labeled as strength-based, device studying with the mixture; these processes have each combined and progressive protocol. The combined intention is to segment: (i) healthful brain tissues, (ii) mind substructures, and (iii) tumor and intra-tumor areas. A revolutionary way is that the difficulty of the technique will increase. Generally, the result of this overview may be divided into 4 predominant regions: (i) the principal challenge in segmenting brain structures, (ii)

segmentation approach trends, (iii) types of brain structures which have been segmented, and the calculation time of brain shape division. Although the late straits in the mind photo divided techniques, numerous challenges will still be alive. CNN is used. We achieved the results and accuracy of tumor and non-tumor cells by testing and validation. The accuracy achievements are shown in figure 3 and figure 4. The final result are shown in figure 5.

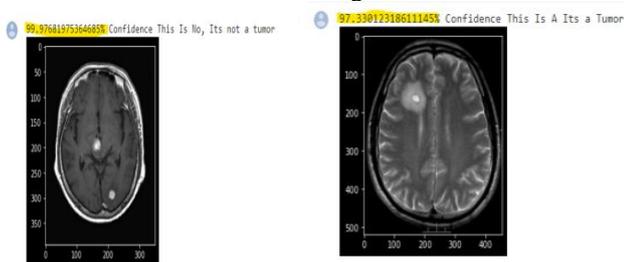


Fig 3: Accuracy achieved

Fig 4: Accuracy achieved

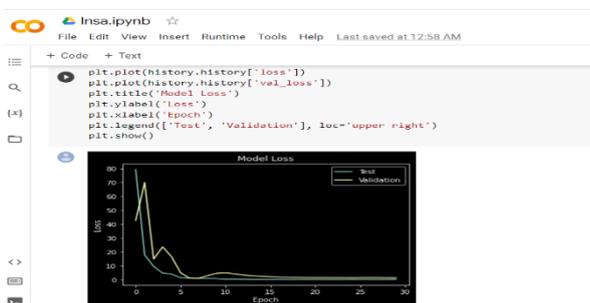


Fig 5: Final result graph

V CONCLUSION

The brain tumor segmentation and category task have great advantages in enhancing the prognosis, managing making plans, and observe about sufferers. By applying diverse strategies, such as traditional picture processing, shallow machine learning technique, and CNN techniques, in dubitable development was done in automated brain tumor segmentation and categorization obligation. But constructing a completely self-sustaining device that may be used on medical flooring continues to be a tough challenge. We collect datasets for MRI images publicly /Internet. We use two models CNN for the prediction of brain tumors to achieve better accuracy. Analyze the area of images and which type of cells is affected and predict brain tumors at an early stage.

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