



# Detecting Alzheimer's Disease using ML



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**Preface**

Medical images such as magnetic resonance imaging and computed tomography are an important way to effectively diagnose human diseases. The traditional method is to perform a manual study of brain tumors based on the image assessment performed a radiologist / doctor, which can lead to an incorrect classification when analyzing a large amount of MRI. To avoid human errors, a semi-automatic intelligent classification system is proposed. One of main source of human loss are brain tumors. If the tumors are discovered rightly at an near the beginning stage, the probability of endurance can enhanced. Magnetic resonance imaging (MRI) is important for identification of AD. Neurologists make a diagnosis through image analysis or signal examination. Experts currently use clinical applications that help them make decisions. This type of application uses computational cleverness. These applications can establish high precision whether a patient is well or has a disease. presently, several computational cleverness technologies are exploit for classification of AD. Magnetic resonance imaging technology is used in learning of person mind. In this process, a classification technology found the support vector machine is projected or functional to the classification of mind images. Therefore, convolutional neural networks of Mean, Std and Homogetiny will be used to extract characteristics of the MRI imagery. The planned process will be base onSVM or will diagnose with MRI as input and classify it as AD or common. In work, we suggested a diagnostic model of binary AD support the extraction of deep features for classification by magnetic resonance imaging. The objective is to acceptably classify patients with AD or not with disease.

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## **Chapter 1**

### **INTRODUCTION OF ALZHEIMER DISEASE**

Medical imaging, such as magnetic resonance imaging and computed tomography, is an important method for successful analysis of human disease. The traditional process is to perform a manual analysis of brain tumors based on radiologist/medical examinations, which can lead to classification errors when investigating multiple types of MRIs. To avoid human error, a semi-automatic filtering system is implemented. One of leading reasons of death is brain tumor. If first signs of a tumor are correct, it may increase your risk of survival. Magnetic Resonance Imaging (MRI) is very important in study of AD. Researchers create tests using image analysis or graphing. Researchers use clinical applications that help them make decisions. This type of application uses computational intelligence. These exercises can clearly determine whether a patient is healthy or ill. Today, a number of spyware technologies are used for the classification of AD. Magnetic Resonance Imaging (MRI) technology is utilized for learning of human mind. In this process, support vector machine (SVM) classification techniques are developed or applied to image classification of the brain. Therefore, a convolution neural network of mean, standard deviation, and homosexuality will be used to remove MRI features. The presentation is initiated by support vector machine (SVM) machine and will use MRI as input to detect MRI and categorize it as AD or normal. In this paper, we advise a fresh AD diagnostic model founded on discovery of deep features for magnetic resonance classification. The aim is to better classify AD patients rather than disease patients.

#### **1.1 Alzheimer Disease**

Alzheimer's disease is named after Dr. Alois Alzheimer's. In 1906, Alzheimer's doctors noticed a change in the mental organization of women with unusual mental illness. Alzheimer's is the most common dementia. It is a life-threatening language, with memory loss and other mental faculties that have a profound effect on daily spying. It accounts for 50% to 70% of demonetization cases. The biggest risk factor identified is age. People with Alzheimer's are over 65 years old. But Alzheimer's is not just a disease of the elderly. Up to 5% of patients with this disorder develop better with Alzheimer's disease (also known as early childhood) and often appear in the 40s and 50s. Over time, Alzheimer's disease

worsens. It is a hereditary disease that slowly deteriorates over the years. In the early stages, memory loss was minimal, but in the later stages people lost capacity to talk and answer to the environment. The development of biomarkers (biological processes in particular) is essential for fast exposure of Alzheimer's disease. (Loves Stone, 2009) How to diagnose Alzheimer's disease depends on ability of physician, the history of the physician, and the ability of the patient to respond to the physician. There are no tests that can diagnose Alzheimer's. The purpose of physical activity is to examine general health and the regulation of conditions that may affect the mind and function. Experts believe that specialist doctors can diagnose Alzheimer's disease with an accuracy rate of over 90%. As more and more test takers, medical and marketing agencies develop; dementia testing is aimed at people directly. The Alzheimer's Association trust that home screenings cannot or should not be replaced by a full-blown doctor's examination. When diagnosing a person with Alzheimer's, your doctor should follow an established diagnosis. Although the results of the interview are not intended to provide a 100% accurate test, all tests performed can cause great mental distress to the patient. Other tests that doctors prescribe are review of medical records in which the doctor investigates the someone being check or their relations to collect data about past illnesses; . He has a general understanding of the issue or feels good; he knows the date, the time or the place or can memorize squat language and pursue instructions or do easy analysis, etc. The (MMSE) is one of the most common exams in mind assessment. At the MMSE, mental health professionals consult patients in a sequence of articles intended to assess different levels of mental health. The highest achieve for the MMSE is 30 points. A score of 20-24 indicates mild dementia, 13-20 indicates mild dementia, and less than 12 indicate severe dementia. On average, the average MMSE for Alzheimer's patients decreases by 2-4 points per year. All of these tests require the expertise of the patient and the physician.

New experimental technologies have modernized the accepting of formation of existing birds. Currently, routine medical examinations for Alzheimer's disease a lot include compositional resection by MRI, or by using Computed Tomography (CT). These images are mainly intended to identify evidence of a tumor, soft or heavy bone, and severe head or spinal cord injury. Researchers are working on examining whether use of magnetic resonance imaging and other methods can be improved to play a direct role in the diagnosis of Alzheimer's disease.

Studies have also exposed that waste of specific brain region, particularly the hippocampus, may be a precursor to Alzheimer's disease. The hippocampus is a major part of human brain and other mammals. It is based on the limbic system and plays a key role in long-term recall



and spatial visitation. However, scientists have yet to agree on a standard value that determines the importance of a specific amount of contraction by any person at any time. Their texture characteristics can be used to identify AD.

The late period of Alzheimer’s disease may also be called “severe”. In this phase show reduce mental capability, total loss of cognitive purpose or conclusion this reason death. Today, approximately 36 million persons are living with Alzheimer’s disease universal and this figure is probable to double increasing to 66 million by 2030 or even to triple to be 115 million by 2050 (Hebert et al., 2013). Indeed, every second somebody in world increase Alzheimer’s infection. The global estimated cost of dementia worldwide is US 600 billion dollars (380 billion e). For instance, in United States, there is an probable 5.5 million persons of all ages have Alzheimer’s infection by 2014. This embrace an probable 6 million public aged 65 or adult, or around 200,000 entity age 65 who have.

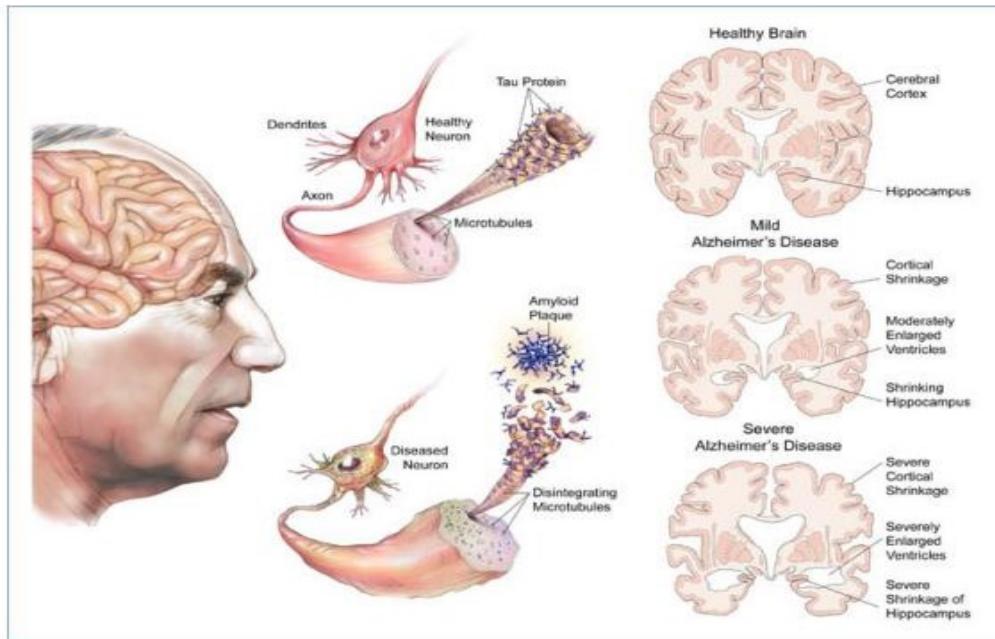


Fig1.1: Cells Degeneration Process and Brain Shrinkage

In the case of Alzheimer’s disease dementia spatio-temporal progression of the disease (By Chess Coach Will Stewart)

## **1.2 Objectives**

The present research intends to find out the most suitable pre-processing technique among the available ones and a new one based on wave atoms shrinkage, which is a multi-resolution technique and is not yet used for medical image de-noising by image processing. The de-noising is extraction of a signal from a mixture of signal and noise.

Also a new algorithm is developed to segment images after pre-processing with good accuracy. The whole system will be used to diagnose the Alzheimer's disease at early stage. This can be used as a biomarker to detect memory related diseases. Since it is automatic, it is user friendly and accurate. After the preparation of MRI using various techniques, the study is estimated in the calculation of accuracy, accuracy, and memory. The number of applications used in the classification is decreasing. The support vector machine is used to categorize the disease. Classes are assessed by calculating accuracy, sensitivity, caution. Develops a systematic, self-regulated approach to assist the neuroscientist in dividing the brain's path to magnetic resonance imaging. This approach has to analyze and integrate human resources at the same time. The method must be successfully performed with magnetic resonance imaging (MR) ambiguity and be able to determine the knowledge of the tumor in the image.

## **1.3 Motivation**

Experts classify the brain as a picture of a tumor or a tumor is a time-consuming process. In addition, disagreements among field experts are irreversible and this is not possible due to the large amount of data needed for statistical significance analysis. It stimulated the theoretical need for imaginative categorization for them. Automatic classification can have a great impact on clinical medicine by reducing the burden of manual labeling by physicians and providing a reliable measurement of the disease. One of the problems with medical imaging is the automatic classification of brain MRI. The purpose of medical image processing is to automatically provide existing tools that are useful for medical groups, but still need the help of healthcare professionals. This is a complex task that involves a variety of disciplines, including pathology, MRI physics, and radiography and image analysis. In this process the introduction is a picture for the brain.

## **1.4 Different Techniques for Ad Detection**

### **1.4.1 Magnetic Resonance Imaging (MRI)**

It provides a non-invasive mechanism for human assessment dissimilar radiography or computed tomography, magnetic resonance imaging does not use ionizing emission or is more effectual in exploratory flexible tissues such mind, spinal cord, muscles, strength and tendons. As a result, MRI is far superior to other imaging tools by providing high-resolution non-invasive images (including chemical information) for most applications. MRI uses the large amount of water available in the body. It is much more flexible than radiographs and CT scans and can generate three-dimensional images in any direction and depth of the human body. Brain images include MRI and computed tomography (CT) of mind structure to rule out tumors or blood spots in mind. Positron emission tomography (PET) scanners can see how certain parts of the brain work or are active. Many scientists are trying to determine if extra mind imaging method can identify reliable signs of before time Alzheimer's disease sufficient to be exploit as a diagnostic tool.

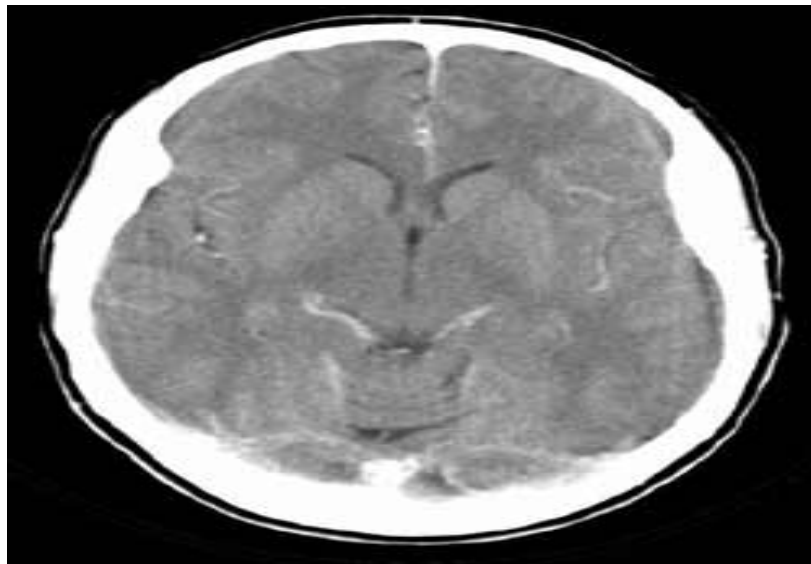


Fig.1.2 MRI Image

It transmits powerful magnetic, radio waves, and computers to invented thorough images of organs, fragile bones, bones, all inside systems. Magnetic resonance imaging can sense mind aberration connected with mild cognitive impairment (MCI) or may exercise calculate patient through MCI may enhance Alzheimer's infection. In premature point of Alzheimer's disease, brain MRI can be quite common. In later stages, MRI may reflect the size of the brain (which has a significant effect on the temporal and parietal regions) of different areas of the brain.

### **1.4.2 CT (Computed Tomography) Imaging of the Head**

CT scans combine specialized X-rays with a powerful computer to generate multiple images or images within the human body. Doctors use brain CT to diagnose and diagnose additional source of dementia, such as brain tumor, subcutaneous hematomas, or blood vessels.

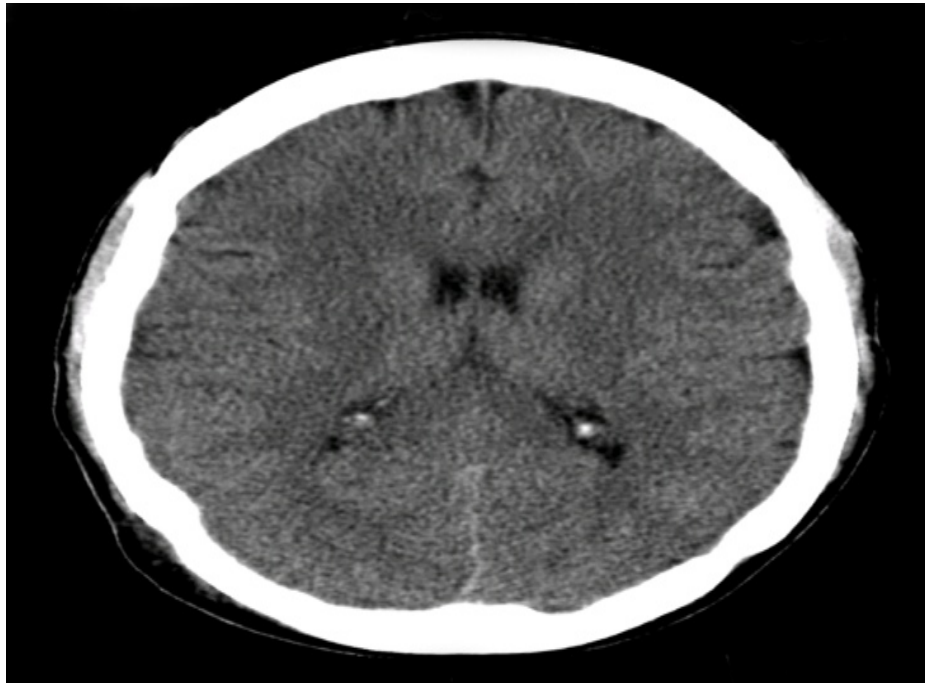


Fig.1.3 CT Image

### **1.4.3 PET and PET/CT of the Head**

Positron release is a diagnostic test that uses tiny amounts of radioactive objects to detect or establish the cruelty of various infections. The combined PET / CT exam combines the PET and CT images with detailed information on the anatomic (CT) structure and function (PET function) of the organs or bandanna. PET / CT scans can help distinguish Alzheimer's disease from extra figure of dementia. To accomplish this, a nuclear medicine trial, also known as a single-photon loading technique (SPECT), was developed.

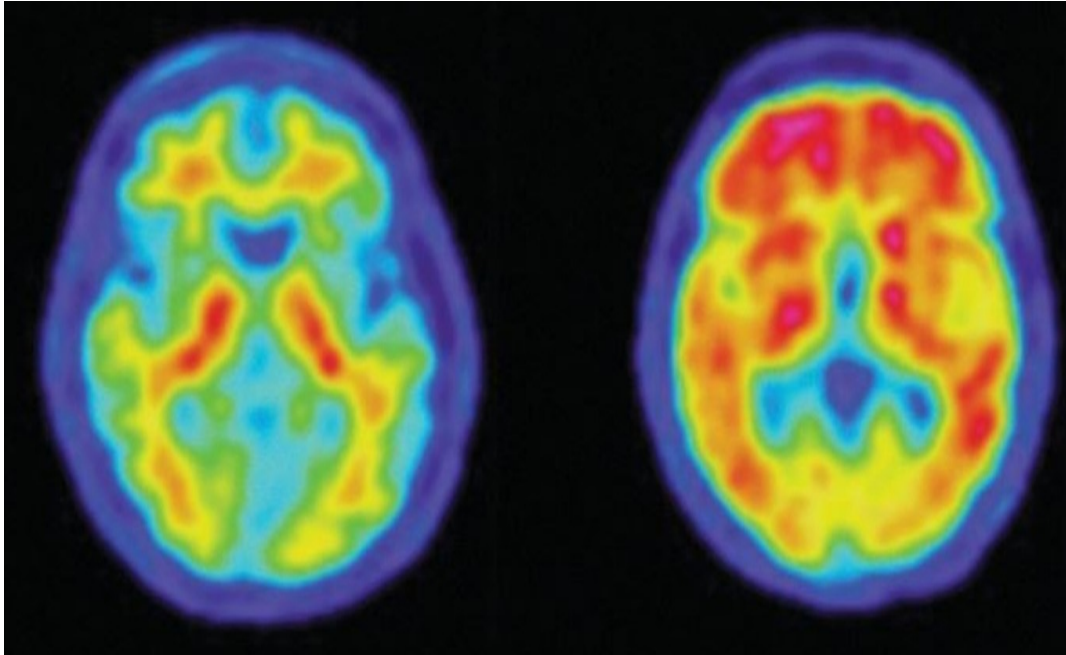


Fig.1. 4 PET and PET/CT of the Head Image

### 1.5 Computer Vision for Medical Imaging Diagnosis

Due to the huge increase in the diversity and number of biomedical image collections or wide range of picture modes currently obtainable, there is required to provide programmed tools to index and manage medical in sequence. The field of computer revelation has aroused greater interest from several investigate community in the field of medical image management. CAD is an interdisciplinary technology that combines essentials of reproduction intellect or computer visualization with radiological or pathological figure procedure. A typical application is to detect tumors. For example, some hospitals use CAD to support X-ray preventive physical exams (diagnosis of breast cancer), colon polyps or lung cancer. In general, computer aided inspection (CAD) systems are limited to structures and marked parts. A computer-assisted diagnostic system (CAD) evaluates visible structures. For example, In mammography, CAD can highlight micro calcified groups and ultra-soft soft tissue structures. These permit radiologists illustrate finale regarding pathological conditions. Another application is CAD, which can quantify, for example, the size of the tumor or the behavior of the tumors during absorption of the contrast medium. Simple computer-assisted classification (CAST) is any more kind of CAD that allows fully automated primary explanation classification studies of certain significant categories, such as pessimistic or

affirmative. CAST is particularly suitable for emergency diagnostic images, which require a rapid diagnosis of serious life-threatening conditions.

### **1.6 Content-Based Image Retrieval (CBIR)**

CBIR is purpose of computer vision techniques to better human picture substance understanding and to index images with minimal human intervention completely regular standard or disease creature mind recognition from MRI is of huge significance for investigate or clinical learning specially in Alzheimer's disease diagnosis application. For this aim, advances in computer vision and growth of medical imaging techniques permit together for revision structural modify in person intelligence and their relationship with clinical diagnosis of AD. Medical sequence Structural Magnetic Resonance Imaging (SMRI) and Diffusion Tensor Imaging (DTI) are used for detecting structural abnormalities of human brain or follow development of mind weaken which is measured as a indication of AD method. Often, clinical diagnosis is based on a arrangement of medical images according to the anatomy of specific ROI (region of interest) known to be occupied in infection rather than the entire brain structure. Sometimes the distinguishing features that would indicate a particular classification are difficult to recognize even by a trained expert. The application of content-based indexing, classification and retrieval techniques in CAD (computer aided design) has obtained increasing research interest by using the visual appearance of MRI tissues. The diagnosis here is based on classification of local individual pattern. Feature vectors extracted describing low level features in an image, is a basis of similarity measurement in a retrieval/classification procedure. Computer vision deals in general with information extraction from images. A variety of illustration quality, such as texture, shape or spatial relationships has been used in other domains, have been adopted in the medical domains these features with little alteration.

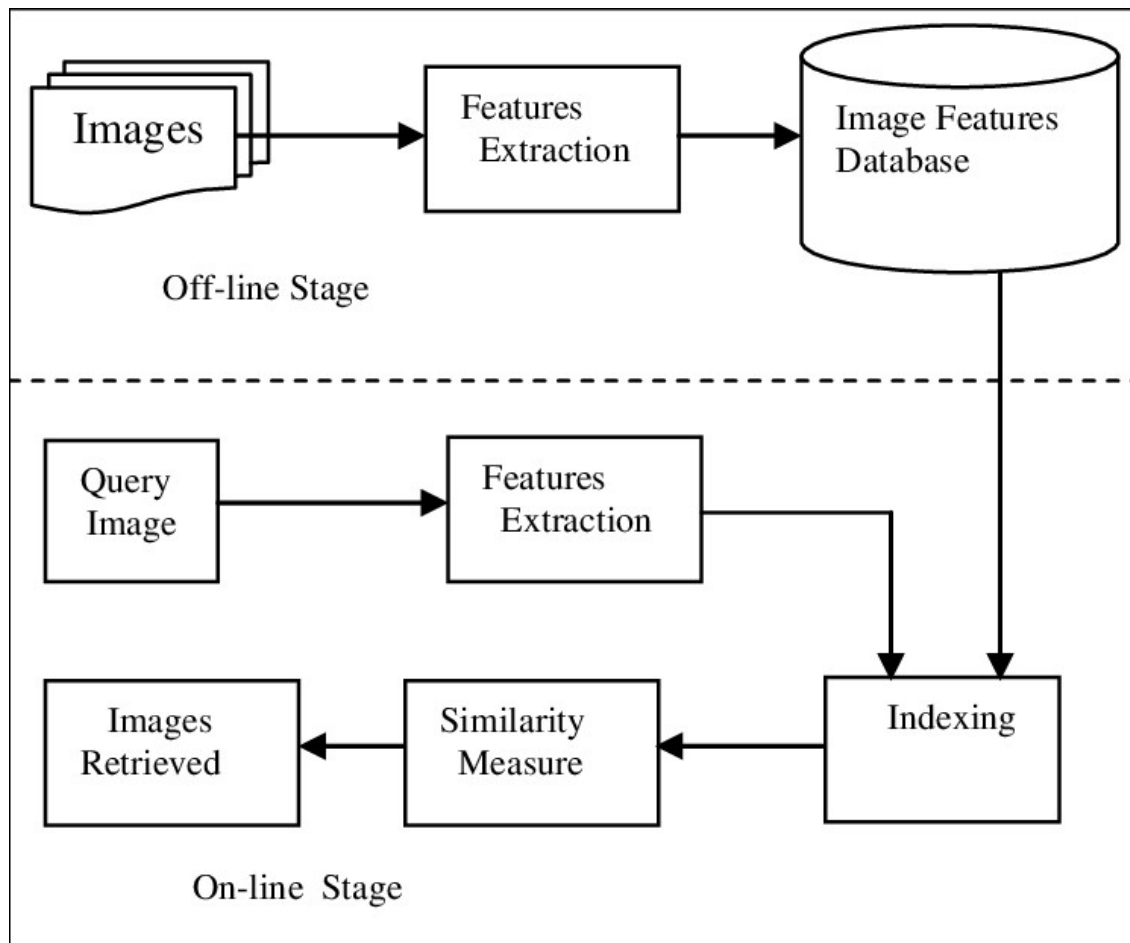


Fig.1.5 Content-Based Image Retrieval (CBIR)

### 1.7 Magnetic Resonance Imaging Theory

Magnetic Resonance Imaging was mainly grow around 1980. MRI is support on phenomenon of Nuclear Magnetic Resonance (NMR), which lead to several Nobel prizes (Geva, 2006). MRI has presented itself as a powerful imaging technique as a way of visualizing detailed structures in-vivo. It is based on magnetic manipulation of protons to acquire images without ionizing radiation. This techniques the patient is placed in an MRI scanner in a muscular magnetic field. This magnetic field reasons hydrogen particle in patient's body to support either in similar or anti-parallel to field.

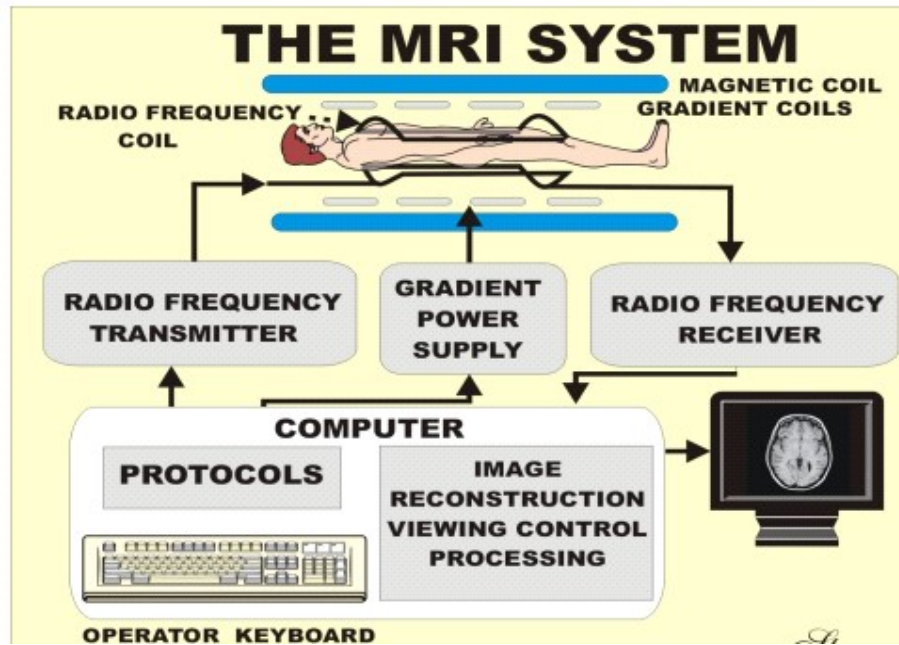


Fig 1.6: Components of Magnetic Resonance Imaging System

Fig 1.6 explains major machinery of Magnetic Resonance Imaging method. Radio-frequency coils in machine emit radio-frequency (RF) pulses causing the proton to spin on its own axis. When RF pulse is turned off, protons go back to being aligned with static the static magnetic field and send electromagnetic energy back to the radio-frequency coils. This magnetic resonance signal is used to produce the 3 Dimensional grey-scale image. The rates of the proton spin relaxation can be in different, depending on the tissue type they are located in. This is how, brain tissues are distinguished as gray and white matter in an MR image.

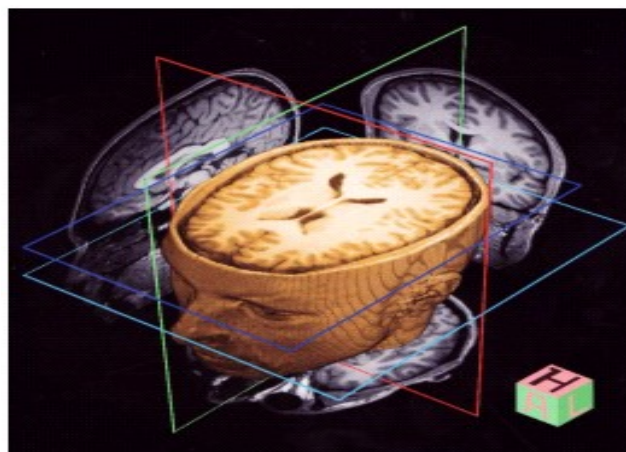


Fig 1.7 T1 (Anatomical) Image Taken of Tim's Brain



## **1.8 AD Approaches**

As with most major chronic illnesses, AD treatment can be divided into three categories: chronic illness, chronic obesity, and chronic disease.

Current therapeutic agents (cholinesterase inhibitors and glutamate antagonists) have symptoms and affect cognitive function. There is still a greater need for treatment than for the treatment of disorders, and these formal therapies will have a greater impact on the cognitive domain as well as other depressive symptoms such as rest, psychosis and sleep.

Most of the ongoing efforts to discover new treatments have focused on the change / progression of the disease. The priority is that if the onset or progression of the disease can continue, it will take longer for people to develop AD. The third category, rehabilitation, is still a hypothetical way. Rehabilitation of the diseased brain and replacement of muscle and spinal cord is too much for AD.

## **1.9 Organization of Dissertation**

The rest of the dissertations divided into 4 chapters.

**Chapter 2:** Chapter two included previous research and literature survey on different Alzheimer disease detection technique and prediction using machine learning and possibilities of AD, machine learning.

**Chapter 3:** gives an outline of aspect of Alzheimer's disease and proposed method for finding of AD followed by image pre-processing .quality extraction, feature detection or disease classification.

**Chapter 4:** The computational characterization of AD protein molecules through structural and sequence analysis and result analysis.

**Chapter 5:** Summaries the findings of research conclusion and explains the scope for further work and limitation.

## Chapter 2

### LITERATURE REVIEW

This chapter describes various existing approaches and techniques to diagnose AD. The diagnosis of AD is a main problem. Automated image based classification for individual patient's plays a pivotal role in computer assisted diagnosis.

Ongoing AD research lays emphasis on identifying biomarkers which best predict future cognitive decline, especially at the earliest stages of disease progression. The expansion of computerized marketing programs based on MRI additional medical imaging technologies has attracted attention clinical drug. It is significant to message these technologies are intended to help clinicians gain additional arithmetical proof in testing and ultimately, hope that these biomarkers can be used as early markers of AD's.

Support Vector Machines (SVM) By Krishna Thulasi[1] - Alzheimer's disease (AD) is one of major disorders of mind, which regularly impairs recollection or thinking ability, or then impairs capability to perform standard functions. This is a good reason in older people. As you get older, dementia becomes more common, but it is less common. AD is a memory muddle in AD, which accounts for 80% of all infections. The three AD stages are mild, moderate, and heavy AD. In the case of mild cognitive impairment (MCI), the loss of critical knowledge alone does not affect everyday life, and mediation is the central step in AD. With many ADs, people cannot work alone and rely on others for care. Support vector machine (SVM) is used to identify and classify MRI of the brain in Alzheimer's disease. The algorithm was qualified or tested use MRI data from the Alzheimer's Disease Neuroimaging Project (ADNI). The information used included magnetic resonance monitoring in approximately 70 patients and 30 continuous monitoring.

In traditional IS methods, users move from view to edge of the area of importance, or move the mouse cursor to limit. As the mouse controller approaches the boundary, the control of the fire is controlled by the boundary. Therefore, when the mouse moves to an area of interest, the active wire moves in that area, resulting in a funny border. Similar to traditional IS methods, in the case of the user-provided EIS, the initial value is selected by region. However, instead of following the parameters of the mouse pointer to the limit, the user selects a set of parameters approximately limit. If the user chooses a point near limit, location for the area of interest is selected and the area of interest is defined. So when choosing a location, it will split. The point

at which the user chooses the EIS can be set close to the limit and should not be set to border. The major benefit of traditional methods defined by IS users is that users do not need proper supervision. The user simply clicks the edge of the area, and the EIS algorithm automatically maps the area about the place of attention. This authority for earlier consumer communication while present limited access support on customer awareness. The main advantage of using HMM to address the problem of fixed mining is that solutions such as the Viterbi algorithm can be found efficiently. The IS approach, where no problem can provide a solution. The success of the calculation is crucial to the proposed EIS algorithm, it aims to provide physician with immediate user interaction.

In this article, we introduce Enhanced Intelligent Scissors (EIS), a new method and fast linking method for medical image sharing. By using local price based on the time involved in the wavelet process, the algorithm is robust to consistency and noise. By presenting the problem as an HMM, the new viterbi algorithm can be used to find and solve the best limits on the points the user chooses. In addition, the new development strategy has been planned to progress presentation of algorithm..

**Advantage:** The major benefit of using the future consumer communication method over traditional IS methods is that the user does not need to carefully trace the limits.

**Disadvantages:**

- IS technique is advanced base on picture gradient, creation it extremely responsive to contrast non-uniformities classically found in medical pictures.

**Multiple Kernel Learning By Michele Donini[2]** - The problem of neuroimaging-based diagnostics, combining different sources (such as MRI images) with clinical data is a difficult task. Compared to using the best resources, simple dating often does not lead to improvement. In this article, we will study the popular dates for Alzheimer's disease Neuroimaging Project (DNA), which includes surveillance and AD activity. We used a new study methodology proposed recently called EasyMKL, which combines multiple parks with workflow options to find the best solution to improve translation. The new EasyMKLFS method improves the b value by eliminating local procedure as healthy as Simple MKL or the latest method. The structure examines waterfalls in the boundary weight chart. We define the water flow based on the irreversible idea of groundwater flow. First, we define similarities of these waterways: they can be distinct in a like way by "talking point" or by "dividing line" separates them. We then prove their sensitivity to the smallest forest through the same theory.

After that, we present a linear time series of calculations. To our knowledge, parallel assets have not been validated in other systems, and the algorithms presented are theoretically and in the most efficient way. Finally, the concept of elucidation is explained in the image distribution, and it is concluded that this method improves the feature of the water-based partitioning of the test image.

In this article, we have introduced the breakdown of waterfalls, which is the concept of water flow in a weight chart. We have been testing inequalities and perspectives of waterfalls: they can be defined by the most obvious dispersal in waterfalls and the rules of waterfalls (boundaries); it resembles the forest with the smallest distribution. Separate. We then presented a linear time series to calculate the basin. To our information, to proposed algorithm is most effective algorithm today. Finally, we illustrate the process of water extraction in image distribution and show that they can improve the quality of contours in water distribution. Among these, we propose a new modification procedure that defines the water supply in a similar way. This redesign allows for the introduction of high-speed snapshots (for example, placing a branch in the middle of a page or inducing a marked tangent), and paves the way for the strategy to work. On the other hand, with this new modification, we can consider the similarities & differences between water tangents tourism paradigms (such as image conversion, image modeling methods). The results of this study include reduction of baseball risk. Therefore, the profile of bowl inherited unique characteristics of test on the plate.

**Advantages:**

- They are corresponding to division persuade by lowestspanning forests
- relation to regional minima.

**Disadvantages:**

- • This includes reducing the value of the map while preserving certain attributes of the route, i.e. keeping the number of elements associated with subspace.

SVM Classification Given By Lilia Lazli[3] –Computer-aided diagnosis (CAD) method that distinguishes patients with dementia, AD or healthy patients. Based on: (1) A hybrid method for the evaluation of white substance, gray matter or cerebrospinal fluid by magnetic resonance (MR) and brain imaging by posterior tomography (labeling). (2) Distinguish between normal Brain and normal image. The starting step consists of three ladder. In addition, the c-mean distribution algorithm (FCM) is worn in the first b-centroid

section. Second, the Cuz grid C grid algorithm (PCM) is used to correct the weak mapping, which uses FCM partitioning to obtain the final image. The last part aims to determine the size of the brain. In the classification process, vector machine (SVM) is used for many tasks. Based on the three FCM methods, the PCM and CAD systems were validated with AD and MRI picture of 45 to 50 healthy individuals. About 55 to 90 years old. Gender, sex, specialty and honesty at work. Speakers (20% noise) had 75% accuracy on MRI and 73% on PET scans, compared with 71%, 70.2%, and 67%.

The idea behind this article is to use mature matrices and drawings for image recognition. All we know is that no one tried to say this. The idea is simple. Each image, a vector shape, was created by removing the maturity of the vector-generated vectors (for classification). In addition, the Haralick masterpiece with 14 statistics can be used to generate new vectors of 14 elements. The proposed method is compared with a four technique called PCA which a general project for reducing the size or shape of the image to obtain statistical models and signal processing. This approach is compared to the popular anti-discrimination approach, which surmount limits of the PCA analysis of concern grammar. Fisher's methodology is to collect picture of categories or divide the images into different categories. The third method is the local binary model, which is a numerical operator that describes the operation of the local space system and image classification. Finally, compare the svm with the Gabor wavelet processing. The edit includes rotating the image through a Gabor filter with tags and programs.

**Advantages:**

- It is obvious from outcome that svm is a healthy analysis for face detection with competitive presentation.

**Disadvantages:**

- Their computational rate in data pre-preparation and change to other spaces.

SVM Classifier Using By Stefano Ginestroni - [4] The potential of drugs that can alter or slow the progression of the disease, early or exact detection of (AD) during neuroimaging information has become progressively more attractive. In this article, our goal is to enhance a slightly successful machine learning solution that uses structured magnetic resonance imaging to identify and distinguish people exaggerated by mild AD. Selected features include unidirectional and bidirectional permutation of subcortical or cortical level, as well as cortical depth or twist of multiple brain region that are identified to be susceptible to AD. In addition, to compare the accuracy of the scheme, several combinations of features

are fed into the (SVM) and the Naive Bayes classifier. The greatest combination of characteristics and M classification schemes, using SVMclassifiers with 86% accuracy ,82% Sensitivity distinguishes patients with mild AD from controls. Although this study is only of a preliminary nature, more efforts are currently being made for automatic feature selection. optimal classifier determination and parameter optimization, but in terms of disease. Our results seem promising in terms of automatic differentiation of phase accuracy; although conventional methods cannot easily distinguish clinical investigations.

The key point of our work that has worked best with the superpixel segmentation algorithm is the algorithm. Most encouraging is the trade-off between quality and execution time. But in the meantime, there are other show lines. In this work, we suggest a novel superpixel distribution found on an existing algorithm, which provides more flexibility. A good balance means that we contribute a lot to our small payments. It is hoped that the proposed new algorithm will be better suited for the workflow of real computers. The first SLIC isolation algorithm is faster than SLIC and runs on traditional desktop CPUs in combination (image size 30 Hz to 481x321). The speed is back at the expense of lower quality calibration. The second proposed algorithm is a mixed region. Based on the subdivision of the river stream, it is approximately 10 milliseconds per image, compared to a superpixel compiled in the SLIC. We have extensively evaluated the impact of proposed changes to contract changes on the various performance indicators.

The SVM class (svm) is based on the concept of the local k-means group. station centers were launched by a photo booth. The local k indicates that each group considers the pixels in their neighborhood. The band's metric is a combination of the weight of the color distance and the space distance of the image plane. The value of the spatial component (the "compactness" factor) affects the consistency of the shape and size of the components produced. Investment can provide good quality and stability, which the latest super pixel distribution technology in C. Although svm belongs to the fastest category of super pixel algorithms (for example, compared to Normalized Cuts), the execution time of 100 milliseconds can be considered a significant loss (for example, compared to waterfall.) In the following IC article we present a method for increasing the speed of SLIC 3 times. The results examined the impact of this change on the sales force between time and execution.

This greatly reduces the number of pixels in picture area. Although the same section of iterations list is encoded in the main, not all versions are updated with each warning. However, if there is a major change in the nearby petals, the petals that have not been upgraded in one drop will become active in the next move. As a result, additional

computational overheads reduce the number of pixels added or removed for each cluster calculation, and this number is estimated in each cluster and its surroundings, before deciding whether to decide update this cluster to the current update. Water distribution is a fast algorithm. However, their size and shape are immobile and have some limitations.

**Advantages:**

- The proposed algorithms improve some performance measures for a large amount, while only slightly decreasing others.

**Disadvantages:**

- There is no clean distinction between an over segmentation algorithm and superpixel segmentations.

Discrete Wavelet Transform (DWT), Support Vector Machine Proposed By Jesia Mathew - [5] An algorithm has been projected for quantitative study of AD, with a focus on perceive diseases that may cause AD in the future. AD is now common and affects the elderly. Disease causes people to lose sight of their daily tasks, and gradually leads to death. There is no cure for the disease, but it is useful if the disease is detected early because it may provide the necessary help. The algorithm uses AD cognitive normalization (CN), and knowledge barrier (tangible) (MCIC) data classification. MCIC may cause AD in the presently stages. The combination of differential waveforms (DWT) and fundamental component analysis (PCA) utilizes vector support (SVM) machines for extraction and classification. Compared to other algorithms, the extraction of multiple functions from DWT and PCA can provide more clarity in the classification process.

This project suggests a mixing process found on Gray Matching Matrix (SVM) to create a dynamic image along the boundary. Since SVM contains information about the frequency distribution and accessibility information of white hair changes, strategies are essential for calculating. Here, the algorithm is designed to slowly build up to manage the limits of explosives. When confirming the information in SVM, the feature can be considered as a threshold for image sharing. The way it was presented was tested by a picture of a carrot fruit deficiency. To show the effectiveness of performance, the tentative outcome was balance with three other production strategies.

For the proposed technique it depends on the size of the edge quite than GLCM quadrant. In future technique, numerical region is submitted to as DWT, and calculation of the color depends on the color value (the first statistical characteristic) of the input image. This change

in entropy-based techniques, in which the calculation area and the calculated calculation are repeated DWT pixel resolution. Therefore, the proposed method, may provide a better representation of the pixel value of the object limit. Going back to the three images, the results of the improved strategies are obtained using the techniques presented, as shown in Figure 1.5. The proposed technique is available only for the upper right white DWT in the DWT calculation area. Next, calculate the average of the grayscale values included in the calculation area. This signifies that the standard value of white hairs 0, 1 is calculated, based on an entrance equal to 0.5. In the past, approaches have made the threshold look like zero. For most blurred images with higher values of DWT and closer to the lower right limit of DWT, the proposed technique may best represent the pixel values of the limit. Therefore, a better fence is also available. In terms of the bidding strategy, most of the DWT components within the quadrant will be included in the proposed technical calculation area. By looking at all the values of white hair in the area, it is possible to obtain a measurement similar to 0.57. By applying this date, small distorted images can be downloaded, as shown in result. For final image, the projected procedure overcomes the challenges of entropy-based techniques. In the case of natural values, the entropy-based strategy contains some DWT elements in the quadrant together; the proposed strategy will include the entire DWT element in its domain. Therefore, the average of the shades is a good assurance, where the input image is divided into split between the bright and dark pixels.

This shows that the method we use provides the best results for the distribution of all test model compared to additional process. However, when the light is well balanced, all methods provide acceptable results. For the unification of images and media, entropy-based methods, Yun Li methods and Otsu methods have opposite results. The entropy-based approach with Yun Li creates a delicate divide as the order also calculates the flaws and complexities created by collecting neutral light. For Otsu's method, the impact of the distribution of objects is poor (the default order), and their origin is not well separated from the history of the image. The Otsu method assumes that the images with the same modifications have a normal distribution. Select the threshold by dividing the image histogram into two categories to emphasize the differences between the categories. Unfortunately, the constant accumulation of light in the test images and the origins of the difficulty has made the problem of Otsu's calculations impossible. For these five images, the results of the separation of the histogram of the Otsu image are exposed in Fig 8. The result also explains our threshold, the entropy-based method. And Yun Li's technique to a fast-paced picture with no complex background,



the Otsu method recognizes the dark side as something, even if it is defective, because the edges of hell are so tight and unrecognizable in the darkThe histogram of the image.

**Advantages:**

- The future method was tested with starfruit fault picture or answer good segmentation in order to classify area ofimperfection on the starfruit skin.

**Disadvantages:**

- It makes poor outcome when production with loud, compound set and downy boundary images.

Computer-Aided Diagnosis (CAD) Support Vector Machines (SVM) Approach  
ByP. Padilla - [6]This article describes the new computer-aided design technology for study of (AD) based on serious malignancy (NMF) and the standard supportive machine (SVM). CAD is designed to analyze and categorize potential brain functions. This purpose, two mind imaging data were select: a document containing photocatalytic imaging (SPECT) and photographic emission tomography (PET), with data available for patients with heart disease. Alzheimer's disease (EA) and clinical control have been used to address it. These data were extracted using the Discriminant Coefficient (FDR) and Non-Negative Matrix Factorization (MFN) to select and extract the most useful features. The SVM-based production class (with some confidence) specializes in the modified NMF response (with smaller dimensions). The proposed NMF-SVM method is capable of producing high affinity for up to 91%, and has a high sensitivity and specificity (over 90%). This article describes and illustrates a new framework for comparing medical image segmentation classifiers with features based on moments and transformations. Use different algorithms (for example, nearest neighbor k (kNN), growth and learning (GAL)) for things like ultrasound bladder (MR) Image sharing. And the Neural Network (ISNN) Segmentation is performed through the application of feature origin methods (called hybrids) such as 2D Permanent Wavelet Transform (2D-CWT), Short Grayscale Histogram (MGH) and One-Time Combined Version 2W-CWT and MGH. Through various embodiments, the results of the study show that kNN performance is greater than ISNN image sharing. During the analysis, the performance of the KNN, of the three previous mining strategies is compared, and qualitative analysis and investigation of the three classes is performed. The results show that the show of the 2D-CWT and hybrid functions is always better than that of the MGH function in all of its modes.

system or system presented meets the need for the best way to meet the timeline and the standards given in the medical imaging section..

The main function of this work is to find class and removal process to attain a successful image sharing. More recently, the "growth and learning" (GAL) and the neural network monitoring and rescue (GMA) and the neural network monitoring feature (ISNN). The neural network and classifier based on SVM 22 are compared to determine which class has the best experience. Similarly, the 23–24 different categories were compared to validate the results. This paper compares and analyzes KNN, GAL and ISNN under MGH,-CWT and Hybrid to discover the greatest mix of class or "schema of exploitation".

GAL results are better than ISNN when evaluate number, time calculation and performance. You can also see from the results that when comparing the size of the account and the performance of the KNN, the reader is improved than the GAL and the GAL is better than the ISNN. The presentation evaluation was carried out in four ways: American warfare. United States, American ghost law, CT scan and magnetic resonance for accurate visualization, images was acquired using an MRI 11 model.

Growing up and learning GAL is like ISNN. The only difference is that the value does not increase if the input vector is equal to the category. However, when we perform the classification results presented in next segment, GAL is superior than ISNN. The average GAL rate is better than ISNN because the nodes are generated through training or use less training time or distribution measure to ISNN.

After the characteristic extraction, the classification procedure will continue. GAL, ISNN and kNN are apply as a class. The classification development is divided into 2 stages,. Training data, assign values during training, and sketch the original image during the test. Therefore, this is the reason for the use of a control class in which experts select training points from each category for the segmentation ear. Here, KNN, GAL and ISNN are classified as race competitors. In this study, we use KNN, GAL, and ISNN as a class and evaluate the distribution of medical images of the tumor, phantom, CT head, and MR head in combination. MGH, 2D-CWT, and some processing procedureIn the literature, ISNN is better than GAL, but based on work, we can finish that GAL is better than ISNN because it requires less training and distribution time than ISNN, the GAL Not Generated The results of the sectional study were calculated from the magnetic resonance image 11. From the results of the study it can be concluded aISNN is better because of kNN. For the partition it is better than the ISNN under MGH, 2D-CWT and mixed-method analysis..

**Advantages:**

- The structure of a movement or system may meet the need to choose the best way to meet the given time frame and the correct standard for distributing medical images..

**Disadvantages:**

- There are a lot of features required for proper maintenance, but these requirements do not meet the level of work required.

Convolutional Neural Network (CNN), Support Vector Machine is used by S. L. Silva- [7] Alzheimer's disease (AD) is a neurodegenerative disorder that can cause cognitive mutilation in patients. Computational science, especially deep learning, has become a influential means for diagnosing AD. In paper, we suggest an AD diagnostic reproduction bottom on the discovery of deep features for magnetic resonance classification. This reproduction plan to order AD and HC (Control of Health)The data used in this project are images of Minimum Intonansononneau (MIRIAD) in Alzheimer's disease to confirm the proposed method. We selected 30 plates from the area in the eyes of the brain. The architecture of the convolutional neural network is constructed in 3 convolutional plates to capture greatest features of the selected region. Next, we take attributes selected in a vector to study or identify pattern using other expert warning strategies. Finally, the data are distributed through a similar, 10-step method, through the use of different markers, vector machines (SVM) and the neighborhoods are used for training purposes. For the above algorithms the correct results are 0.8832, 0.9607 and 0.8745. Based on comparative studies from other publications, we can demonstrate the effectiveness and reliability of the Alzheimer's disease screening model.

This system introduces a new method that uses the world's best algorithm for converting metric masses into magnetic resonance imaging with multiple comparisons of cancer distribution. To address the issue of limiting differences and high variability in the egg tumor section, a series of measurements from multiple MRI data were first extracted, and we use color mapping to reveal spatiotemporal information not found in the intensified MRI image. . Next, we use an effective metric algorithm with multiple comparisons in the MRI data to create a metric field that captures the world's best labels on uncontrolled systems. Finally, we use a tree-based method to decrease the number of possible markers for heart transplantation. Experiments using real clinical data have shown encouraging results. The method is applicable to a variety of medical methods and campaigns.

The system is based on the direct grouping of the available labels (values or active features) of the medical image data, and the use of the best timber chart construction to detect a world-class brand, but the former focuses on the traditional. Compared to previous work (such as active plates, height, and machine learning), our work on the silent distribution of wood metallic images focused primarily on MRI brain tumors with comparing multiple watches. Our experiments made use of the division of the liver tumor and compared it with the latest technology. The proposed method is also applicable to a range of imaging modalities, such as MRI with contrast therapy, CT scans. To find the limits of tumor distribution in tumor cells through a stable extension of the large-scale differences in the incidence of tumor growth. we propose new methods and strategies for capturing them, and introducing detailed dendrogram (TM) imaging. Algorithm finally, we propose a tree-based method for interpreting the TM algorithm for the distribution of MRI brain tumors with different MRI levels. Here we describe the general protocol for magnetic resonance with multiphase coupling. In multi-resonance magnetic resonance imaging with opposing agents, critics (patients or animals) are often injected with opposite ends, usually G-DTPA (Gd). Collect a T1-weighted magnetic field approximately seconds before injection (before angiography) and time points after injection (end of arterial & portal & pause phase) to obtain baseline or before MRI scans, and when to use interval scanners. When the contralateral operator is repairing the vasculature, the time-dependent contraction of the voxel can be visualized, indicating the MRI scan.

The problem of the partitioning of the liver can be explained as an example for metal markers. In particular, an example from a common spatial unit with observations as a parameter, we need to label new voxels. The new mark is the height of white hair, and the process of building additional trees will result in the accurate segmentation of brain tumors on MRI that is in many respects. Each new value should be close to that depending on the adjustment of the magnetic volume, and the value of the approximate value is the same. In recent times, some literature has talked about the cohesion of interdependent groups. They are all based on iterative techniques, and there are no obvious limitations. In this paper, we propose to use a tree-shaped tangent to calculate the lowest energy production solution for the distribution of the interior of the egg in contrast MRI with multiple agreements.

Evaluate the method for adjusting the method (DPM + TM) with the size of the expanded brain and comparing it with the division results (DsPM + K-mean) in the same scenario. We used our proposed method to distribute liver tumors in a variety of 3D MRI databases and compared them to rabbit lips several times, showing a correlation between our method and gold standard. compared to the same set of ads. . In this paper, we suggest a new technique

that utilizes active-motion imaging and interferes with metallic imaging with tree-cutting for the distribution of liver tumors on MRI dates. the abdomen, and apply the results. Apply to the final method. This method has high efficiency and greatly improves the metallic markings on the wood. In imaging and measuring the size of a tumor it is better than the usual method and avoids iterative methods that require a lot of calculations. There are some interesting comments on future work on this topic. In order to learn the structure of a good tree, such as the optimum speed of a tree and its rules, the distribution of each part of a publication will improve. To find the best partitioning of the brain tumor from the first section of the algorithm, the active learning method can be used to determine the best scan speed in the field of scanning and the depth of deceleration.

**Advantages:**

- The presentation method can be applied to modern images with multiple images, such as MRI, CT perfusion without confirmation.

**Disadvantages:**

- Not providing the best labels in the world.

Fuzzy-C Means By Kai Xiao et.al. - [8]Testing for brain tumors remains a challenge in the field of computer-assisted testing. In magnetic resonance imaging (MRI), the conclusion between long-term changes and the occurrence of a tumor is useful for monitoring and predicting brain tumor. In order to obtain data on later ventricular transplantation for subsequent analysis and statistical analysis, this paper proposes a new method for analyzing ventricular changes. First, in this way, the lateral ventricle is split, the pixels in the border are subtracted, and the non-linear aperture (TPS) is used to create more model images for each particular case. By comparing the application and implementation of neural network (RBF-NN) and blood (trans) neural network (RT) to TPS extraction from bone to ventricular boundaries, iterative rotation is used to find the best image orientation. The theory facilitates the final step of the transformation analysis, which investigates the displacement holes based on the displacement values obtained from the observed model and the diagnostic TPS of the later ventricles. Compared to many cases, our new outcome explains that project is effectual and relevant for analyzing ventricular manipulation and predicting tumor location. Instead, comparative results show that using RT on a sacrifice is an effective way to find the best path to results.

In this document, make sure to use TPS to find the areas of greatest migration. In our previous work, the lateral ventricular limit was obtained by measuring the lateral ventricular distribution by means of a Fuzzy-C refraction assay. Due to differences in age, gender, human brain, and even the number of combined MRI images, although similar to each other, the image may not look the same diagnostics and their temple image. Therefore, the mixed image should be used as a new converged model between the model image and the diagnostic image to improve the accuracy of subsequent processing. Here, TPS-based methods are used to generate more accurate convergence results. Finding a path between the diagnostic image and the MR template is important for generating accurate results in analyzing TPS transmission. Therefore, it is important to adjust the path of the MRI diagnostic image based on association amid diagnostic picture temple image through TPS analysis. As the TPS output data is based on the offset between the diagnostic chamber and the ventricle template, the TPS offset value can also be used as a metric error that is used in game analysis best competition between template and diagnostic images. Therefore, in our initial transfer method, the separation information of the TPS analysis was used as a metric channel through the neural network (RBF-NN). The system shows that the optical speed is at least as fast as the TPS and RBF-NN.

It can be seen that although the model parameters and the diagnostic test sheets are in good agreement, the physical limitations of the developmental approach and their injuries through TPS and RBF-NN methods and RT methods are within the range. Simple. The two methods are different. Although there is a 15-degree difference in the true diagnosis using these two correction methods, there will be no significant change in the final offset position of the TPS. This show that even with different redirection algorithms, the TPS production algorithm can provide long-term results for predetermined bowel positions, and the results between them may be slightly different. Based on the proportions of time used in the two correction methods, it is estimated that the TPS and RBF-NN processing is more than 100 times the RT-correction method. This is mainly because when dividing the address with the least TPS offset, it needs to be calculated in a simple calculation for each diagnostic guide for TPS analysis. Therefore, based on the present experimental results, the RT methodology seems to be a better method because it is more efficient and can provide more accurate results.

This study used a three-dimensional thin layer (TPS). First and foremost, this is a conversion method to restore interpolated images between standard and diagnostic images, reducing their inconsistencies in a limited way. Since the neural network (RBF-NN) is a neural network, it can be used to evaluate all functions when trained as a fault-based sensor, thus minimizing

output. The value of the network can be used to indicate the best address. As you can see, the network can use the following methods to find the best address

**Advantages:**

- It is known that using Radon transform (RT) in bone is more effective, while maintaining similar results.

**Disadvantages:**

- It is not able to be fully solved due to the nature of MR images retrieval.

Semi-Automatic Segmentation Used By Azzeddine et.al. - [9] New insights based on the semi-automatic distribution of active contour models and multiple sclerosis (MS) lesions in the area of interest (RIO) in magnetic resonance imaging (MRI) are presented. Proper segmentation of trust is important for at least three types of practice applications: drug testing, treatment decisions and patient follow-up. The distribution of books on MS lesions in the MRI brain is often preferred by skilled researchers. However, manual partitioning is difficult to replicate and can be expensive and time consuming in the eyes of large MRI data. On the other hand, the process of self-reliance is much faster and may result in reproducible results. However, the results of the divisions produced by these expeditions do not correspond to the ground-truth process provided by scholars. In this paper, we propose a modern Active Contour model based on MS errors, which combines expert knowledge and cost estimation to produce more reliable MS distribution results. Specifically, we choose rigorous RIOs to circumvent their MS lesions and to maintain healthy white skin (WM). Together with the statistical properties of these two categories, it is recommended to obtain genetic features consistent with healthy lesions and MS. The results chart shows a significant improvement over the proposed model.

These paper propose to a semi-automatic computational approach based on a combination of multiple parameters (based on the modified Dirichlet process) and a regional-based model. By combining divisions and classifications, we can take advantage of the characteristics of a given region that provides statistics in a region, defines its overall shape and defines its boundaries. The method we use has the following advantages. First, we rely on mutation sequencing and multiple classification methods to provide MRI scans except for sequential complexity. Second, we have put together a series of features to guide active sharing and practice of mistakes. We counted the Bhattacharyya distance to distinguish the fault. Third,

our approach may be tolerant and applicable to other treatments. Fourth, similar to the other methods, this method is semi-automatic and can be fully automated due to the use of probabilistic brain atlas. Due to the difficulty in finding MS in this field, we use atlas data to determine the cerebellum. Finally, our algorithm results in soft categorization at different stages of MS disease, not just the results. Because prospective errors can vary widely among professionals, this feature may be valuable for clinical analysis.

Next, each PD image is subscribed to the target image, and the registers are redirected to create a new target image. The process was repeated five times, but after further notification, no significant changes were found in the target audience. Permission registration means the difference in area lengths as similarity in assessing the difference between the target image or registered image. Reducing intensity is part of registration process, so all of the images shared share a picture of the latest PD model average. Two representative images from the average PD image are shown in FIG. 1. Using the ROI lesion as a mask, create a visual search of the lesion for each patient, then add it to the temple image coordinate, and use the spreadsheet changes similar to when writing PD images to a template. Convergent binary images are then added, separated by the number of patients in sample, and then scaled to 256x256x256 pixels to obtain a (usually right) lesion per anatomic location. This interchange offers additional spatial convenience, so a small sample of patients can be used to build a map often. In addition, due to space constraints, there are inherent flaws that will occur after the template installation will have less impact..

**Advantages:**

- It is fairly stable for the segmentation of MS lesion and no image-atlas registration is needed.

**Disadvantages:**

- MS lesions are not randomly distributed throughout the bra



Table 2.1 Summary of Alzheimer Disease

<b>Author</b>	<b>Classifier</b>	<b>Advantage</b>	<b>Disadvantages:</b>
<b>Krishna Thulasi et.al</b>	<b>Support Vector Machines</b>	The main benefit of using proposed user interaction method over traditional IS methods is that the user does not need to carefully trace the limits	IS methods is refined based on image gradients, make it highly sensitive to contrast non-uniformities normally found in medical images.
<b>Michele Donini et.al</b>	IS methods is refined based on image gradients, making it highly sensitive to contrast non-uniformities normally found in medical images.	They are equivalent to the separations induced by minimum spanning forests relative to the regional minima.	It consists of lowering the values of a map while preserving some topological properties, namely, the digit of connected components of each lower cross-section.
<b>Lilia Lazli et.al</b>	<b>SVM Classification</b>	It is obvious from the results that the svm is a robust method for face recognition with competitive performance.	Their computational cost in data pre-preparation and transformation to other spaces.
<b>Stefano Ginestroni et.al</b>	<b>SVM Classifier</b>	The proposed algorithms improve some performance measures for a large amount, while only slightly decreasing others.	There is no clean distinction between an over segmentation algorithm and superpixel segmentations
<b>Jesia Mathew et.al</b>	<b>Support Vector Machine</b>	The proposed technique was tested	It produce poor result when dealing with

		with starfruit image and result good segmentation in order to identify the area of the defect on the star fruit skin.	loud, difficult background and fuzzy boundary images
<b>Padilla et.al</b>	<b>Computer-Aided Diagnosis (CAD), Support Vector Machines (SVM)</b>	The structure of a movement or system may meet the need to choose the best way to meet the given time frame and the correct standard for distributing medical images..	There are a lot of features required for proper maintenance, but these requirements do not meet the level of work required.
<b>S. L. Silva et.al</b>	<b>Neural Network (CNN),</b>	The presentation method can be applied to modern images with multiple images, such as MRI, CT perfusion without confirmation.	Not providing the best labels in the world.
<b>Kai Xiao et.al</b>	<b>Fuzzy-C Means</b>	It is known that using Radon transform (RT) in bone is more effective, while maintaining similar results.	It is not able to be fully solved due to the nature of MR images retrieval
<b>ByAzzeddine et.al</b>	<b>Semi-Automatic Segmentation</b>	It is fairly stable for the segmentation of MS lesion and no image-atlas registration is needed.	

## Chapter 3

### METHODOLOGY

This chapter presents several experiments in vivo and in vitro to evaluate the memory enhancement properties of S. Album. The Animal Ethics Committee was approved by the CPCSEA (approval number 1143/365) for use in animal research. White albino mice were purchased from the King Chennai Institute. Mice 4-6 weeks old were selected for study and divided into 7 groups. With the exception of the sixth group, all animals were injected with beta amyloid (a neurotoxin) through the ICV route. The animals were treated with the corresponding drugs and / or compositions for 2 months. Animal behavior was recorded in two compartments in the maze above, the passive avoidance device Hebbs William Maze. Compared to other groups, the memory of animals treated with streptococcal extracts improved significantly. To better understand its mechanism of action, these medications were sacrificed, including the destruction of streptococcal extracts, the brain was extracted and enzymatic activities such as acetylcholinesterase and antioxidant properties were studied. The results showed that in the combined drug treatment of donepezil and S. album, the activity of acetylcholinesterase increased significantly (the value of OD 150 was 3485), followed by the S. album group (value of OD 3220). These values are very significant compared to the control group (OD value of 525). Brain histology proves it further. It is worth noting that the combined dose also showed symptoms of nerve regeneration and the formation of neurofibrillar nodules. Acute toxicity studies were performed at different dose levels and LD50 values were determined. Sublethal doses are used for further studies. The results of previous experiments clearly show that S. aureus extract has an improved effect on donepezil. Even with a lower dose of donepezil, S. Photo albums also have the effect of significantly increasing activity. Histopathology reports support this..

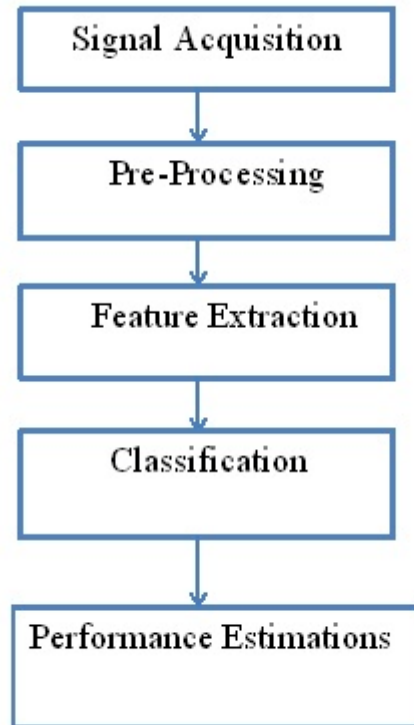


Fig.3.1 Conceptual Diagram

### 3.1 System Design

A tumor is the abandoned enlargement of cancer cells in part of body. Different types of tumors have different characteristics and different treatment methods. Currently, mind tumors are classify into main brain tumors or metastatic brain tumors. The first originates in mind or tends to remain in the mind, while the second originates in cancers in other parts of body and spreads to mind. Segmentation of the mind tumor is key procedures in surgical planning and treatment. Segmentation of brain tumors by magnetic resonance imaging has been a research point. Brain tumors can be of various sizes and shapes and can appear in different places. Changes in tumor intensity in brain MRI Q1 make automatic tumor segmentation extremely challenging. Various intensity-based techniques have been proposed to segment tumors in MRI. Texture is one of the most popular features in the classification and recovery of images. A tumor is an unusual increase of tissue. Brain tumors are a large number of unnecessary cells that grow in the brain or central spinal canal. Brain cancer can be counted as one of the most deadly and intractable diseases.

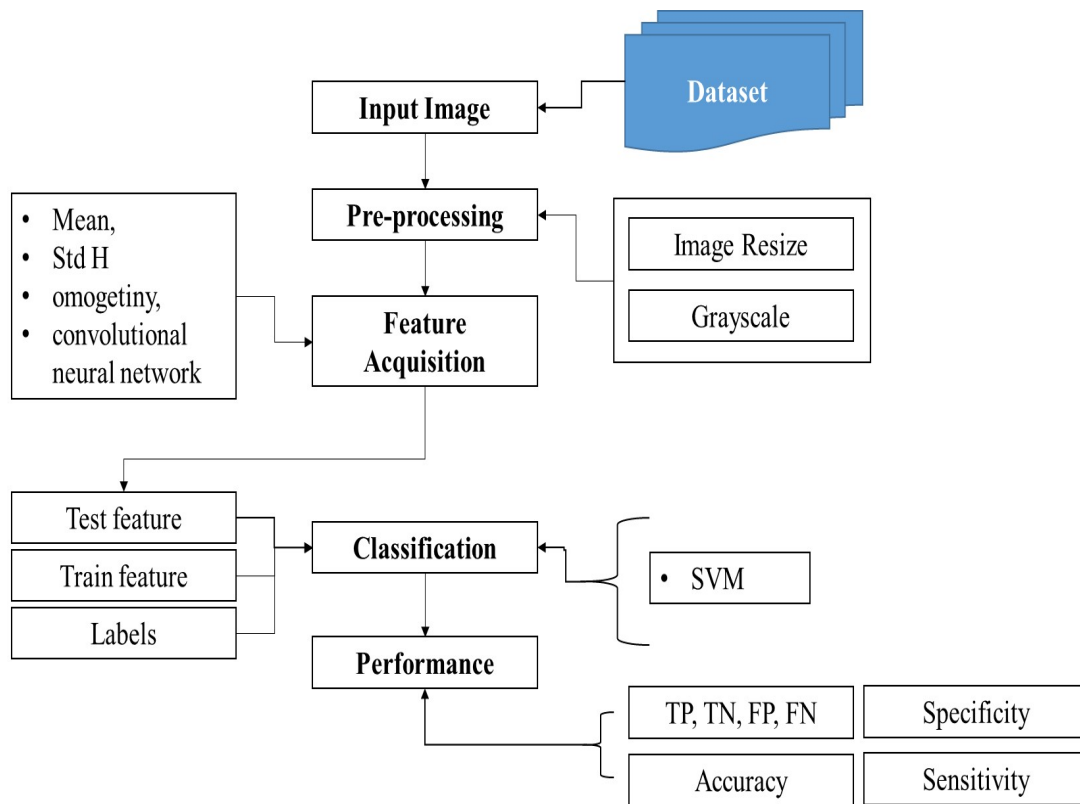


Fig. 3.2 Proposed Flow Diagram

This makes MRI memorable as a good source for researching, identifying and identifying potential areas of brain disease. Nearly all current analytic process are based on human interpretation of MRI scans. Of course, this increases the likelihood of false positives and identifying brain muscles. On the other hand, the application of digital image processing can ensure that the tumor is seen as fast and fast. Segmentation is one of the most effective methods for obtaining complex medical picture information and has wide application in the field of medicine. The idea of image segmentation is to partpictureinto specific and complete area, so that each section of the area of interest is continuous and the segmentation in the area is the same as the predefined standard. The cause of most cases is unknown. Other risk factors include: multiple genetic syndromes such as neurofibromatosis and vinyl chloride transmission, Epstein-Barr virus, and reduced activity.

### 3.2 System Module

1. Signal Acquisition
2. Pre-processing
3. Feature Extraction
4. Classification
5. Performance Estimations

Image processing is a practical way of presenting information stored in pixel form. This investigate work is connected the valuemeasurement of orange fruits. There are several stages in the assessment of multiple quality colors, such as pre-preparation, elimination, or classification. In the previous method, a naive Bayes class was used for the quality assessment. A simple Bayesian classification method is not very accurate in superiorityjudgment. In this research employment, the Naive Bayesian arrangement method is replace by SVM classification for worth assessment. The GLCM algorithm is used featureorigin with a single-part distribution. This approach is compared to existing methods of straightforward, time-consuming, cautious and specific. Compared to the methods for assessing the quality of a fruit juice, this method has more caution, sensitivity, sensitivity and shorter duration.

This chapter discussing about the framework design of “Content-based image retrieval” using “GLCM” (Gray Level Co-event Matrix) method. “CBIR” purpose of computer-vision scheme to retrieve the image content. It means the problem is to search the digital image from large datasets. Therefore this chapter illustrates the detail study of proposed framework design which includes proposed “CBIR” system diagram followed by algorithms flow-chart and result snapshots. The implementation of GLCM is carried out in a numerical computing environment where trial outcomes prove that the proposed GLCM excel improved performance in conditions of accuracy and recall as contrast to the conventional techniques.

Form recovery: Search all forms in a large form database that is generally similar to the form of the query. Generally, determine all forms within a given distance from a query (decision problem), or determine the first forms with the smallest distance (calculation problem). If the database is large, calculating the similarity between the query and the shape of each database may not be feasible. Index structures can often use some form of triangular inequality property to help exclude most of the database from consideration in the early stages of the search.

Recognition and classification of forms: determine if a given form is close enough to the model (decision problem), or determine the most similar of the k class representatives (k calculation problems). Alignment and registration of forms: Transform one form to match totally or partially with another form (optimization problem).

Approximation and simplification of forms: construct the form of fewer elements (points, line segments, triangles, etc.), which remains similar to the original form. The deployment of different optimal methods subjected to improve the coefficient approximation with respect to maximal dissimilarity. However, the smallest value of dissimilarity provides the maximal number of elements.

Segmentation: the local segmentation technique includes multiple stages, such as background Removing, removing the original screenshot, editing the image, reducing the image, modifying the image and removing the screenshot from the mouth to detect defects in the gray color image. In this research work, the disease will be classified using SVM class technology, and the SVM class will find the contaminated site and the percentage of the contaminated site.

### **3.3 Gray Level Image Binarization**

The technique used to improve raw or unstructured images is called picture processing. Few example of image processing include HIS / HSV color change, histogram similarity or color diversity. For illusion dispensation, use RGB or HSV color coding. HSV color gap translates to shade or gloom in red, green, and blue areas. This color space translates color through price, hue, and saturation. The equilibrium of the histogram is considered non-linear. The similarity of the histories involves the distribution of pixels, so that the number of pixels is equal to the values in the matrix.

### **3.4 Texture Feature Using GLCM Algorithm**

characteristic extraction is 3<sup>rd</sup>stage. This step, the GLCM algorithm performs feature extraction on computed tomography images, and the GLCM algorithm is functional to element extraction. The GLCM algorithm will mine the texture characteristics of input image. Calculate selected features. This calculation uses only the values in GLCM.

Intensity: intensity can be defined as a quantifiable number of repeating ranges of pixel pairs.

Intensity is a parameter that measures the similarity of images.

$$\mathbf{Intensity} = \sum_{i,j=0}^{N-1} (P_{ij})^2$$

Eq.3.1

**Entropy:** Entropy is considered to typify randomness of the textural image and is defined as

$$Entropy = \sum_{i,j=0}^{N-1} -\ln(P_{ij})P_{ij}$$

Eq.3.2

Where

$P_{ij}$  -Element  $i,j$  of the normalized symmetrical GLCM

$N$ - Number of gray levels in the image as particular numeral of levels in under Quantization on the GLCM texture page

### 3.5 Gray Level Co-Occurrence Matrix

In 1973, Haralick introduced a method to analyze texture, which is a statistical method of the second order. Haralick offered two steps to get the writing style. The first step is to calculate the co-occurrence matrix, and the 2<sup>nd</sup> step is to compute features of the co-occurrence matrix. From biomedical technology to advanced technology, this technology is useful for various applications in image analysis..

### 3.6 Working of GLCM

The critical principle of the GLCM texture is to consider the relationship among 2 adjacent pixels in a 2<sup>nd</sup> order texture. The gray value relationship in the image is converted into a matching matrix space using a given mask (such as 3, 5, 7, etc.). In the conversion of image space to matching matrix space, neighboring pixels can be used in one or some of the eight defined addresses; These pixels can be used between two pixels. Generally, four directions are first considered, such as 0°, 45°, 90° and 135°, and their inverse directions (negative addresses) can also be considered. There is information about pixel locations with similar grayscale values. Every part (i, j) in the GLCM determines the number of times a value is represented by the gray value close to the j-pixel. In Figure 4.8, calculations are performed where the element (1,1) in the GLCM has a value of 1 because there is only one example



Figure 103 that has a pixel value. Two adjacent vertices 1 and 1. (1, 2) In GLCM the value of 2 is significant because there are two cases in the image, where the relative pixel density is 1 and 2 each. Figure 4.10 Creating GLCM

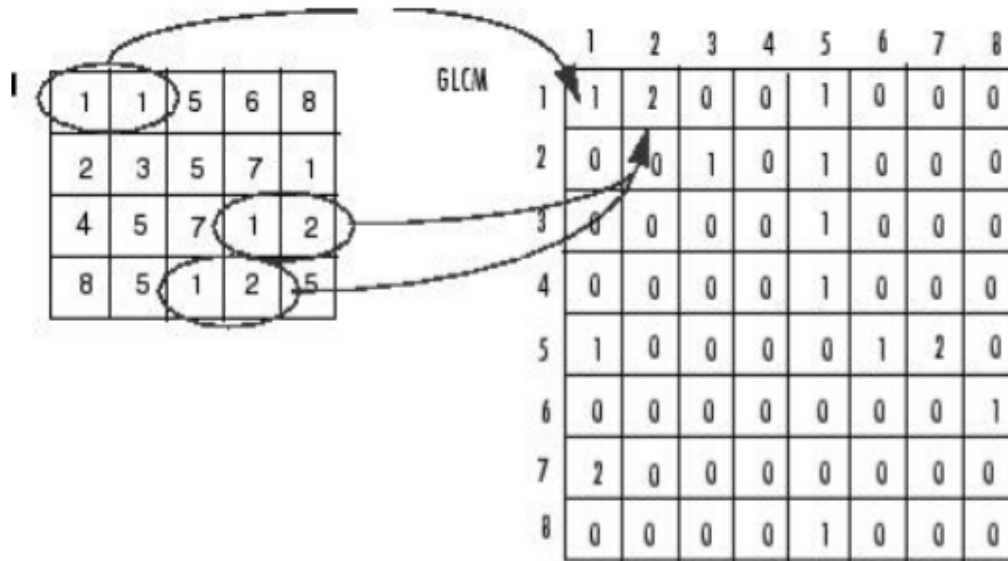


Fig 3.3 Creation of GLCM from image matrix

The element (1, 2) in GLCM contain the value 2 because there are 2 instances in the image, where two parallel neighboring pixels have values 1 and 2 respectively. The GLCM array have been extract for the images in the input data set. Once the GLCM is calculated, the texture characteristics of the image will be extracted continuously.

### 3.7 Classifier

Tree studies use decision trees (as a preliminary model) from a project study to draw conclusions about project values. It is one of the oldest models use in statistics, data removal and engine learning. A model of a tree that can use average values is called clustering, and a tree model that allows variable variables to use constant ideals (usually real numbers) is called a deterioration tree..

### 3.8 Support Vector Machine

In machine learning, vector machines (SVMs, supporting vector networks) are also education models with linked learning accounts that examine data for categorization and redistribution. When providing training examples, each training instance is noted for one or both of these

categories, and the SVM training algorithm will build a mold that provides new example for type or another category, making it. Line drawer.Non-probabilistic binaries (although there are problems with this method, such as the use of SVM's Platt symbol in facilitating probabilistic classification). The SVM model represents the time as points in space and shows them to establish the distance between the samples in each category as much as possible. Then, put a new example in the same section and predict that they decrease into a category based on the length of the hole they fall into.

In addition to doing line categorization, SVM can also use basic techniques called successful in performing non-linear categorization, by mapping through high-speed web space. If the data is not noted, then it is not possible to study controls, and the controls require a learning method that attempts to extract data in a group and then assign new data to these groups. The common selling point for vector machine optimization is called support vector programming and is commonly used in industrial applications. It can be used in industrial applications where no data is specified or some data is classified as a decent precursor..

## **Chapter 4**

### **EXPERIMENTAL AND RESULT DISCUSSION**

The main principle of future systems is to divide the MR images into normal images and abnormal images. Add non-standard tissue images to the classification for both types of low quality and advanced gliomas. Next to the main image, the MR image is preprocessed, such as grayscale conversion, filtering, image enhancement, to create an image that can be used in subsequent steps. The steps we use in the proposed method are shown in Figure 1. For segmentation, we use grouping of k-means to segment the image and find tumor regions. The segmented image is used to extract categories. For this reason, in addition to BWT, GLCM is also used, and due to noise, irrelevant or deceptive richness of the skin, a selection of features must be made. By eliminating these factors, learning from data technology can be very beneficial. The classification of characteristics can be considered as one of the most important problems in field of mechanism learning. The most efficient method of selection of characteristics Fish of generation of characteristics is used to improve the representation of the model, but it is also convenient to verify the results. Finally, use maintains vector machines to organize the descriptions as usual or unusual (lower high-grade gliomas). Keep the edges of the image while eliminating noise. Picture improvement is the development of adjust an image to make it more suitable for later analysis. In the planning method, equalization of the adaptive histogram is performed to improve the similarity of the images.

#### **4.1 Gray Scale Conversions**

In shooting, calculation and colorimetry, a grayscale or grayscale picture is an image in which value of each pixel is a unique example that represents just the total of light that is, it only carries passion information.

This type of picture, also call black and white or colorless, consists only of gray shadows, from the weakest black to the strongest white. Appendix A.1 shows some images of the MRI data set.

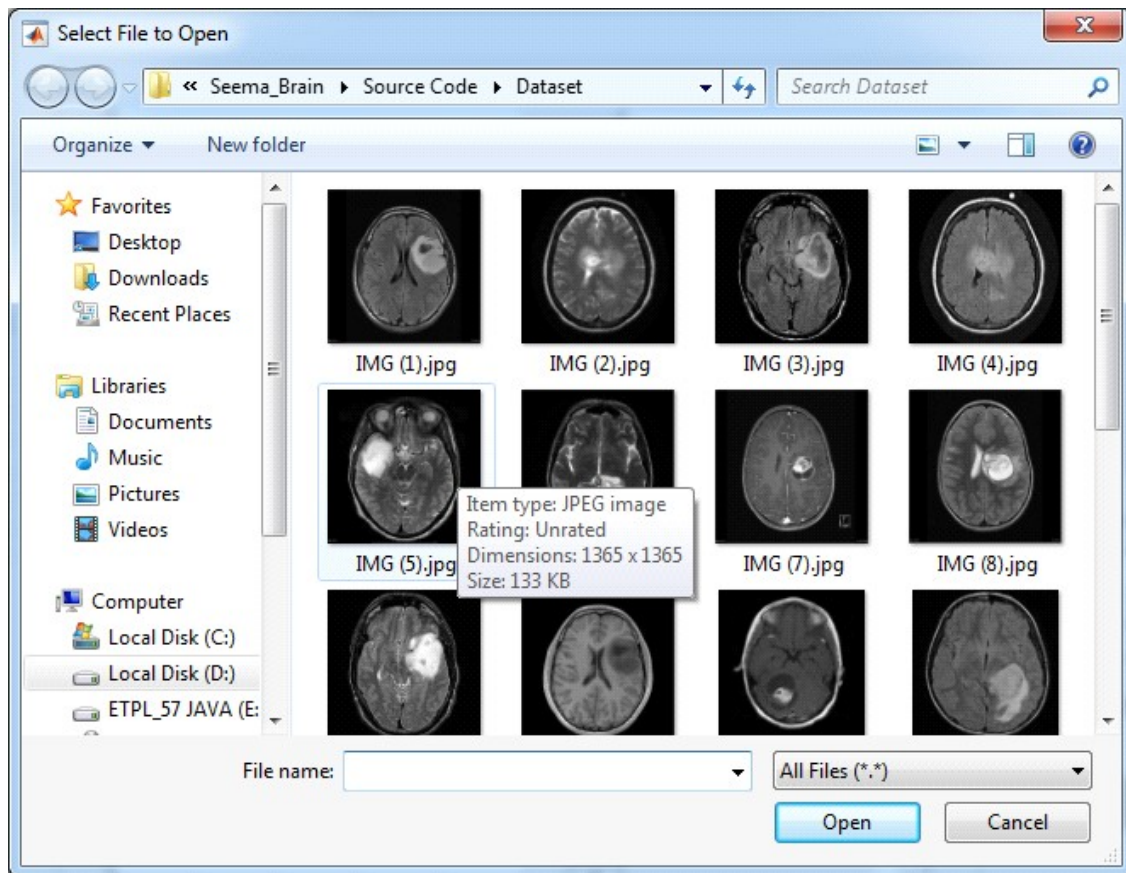


Fig 4.1 MRI Data Sets

## 4.2 Input Image

- The picture is a rectangular range of values (pixels). Every pixel represent a measure of certain scene attributes measured in a limited area.
- This attribute can have lots of things, but we generally calculate the normalintensity (one value) or brightness (three values) of the picture clean by red, green and blue filters.
- These ideals are generally represented by eight-bit integers and provide 256 brightness levels.
- We discuss the decision of the image: this is defined by the figure of pixels or the amount of brightness ideals. Extraction, normalization of intensity, etc. directly affect the outcome of brain tumor segmentation.
- Some input picture is shown in Appendix A.3.

### 4.3 Pre-Processing

Before introducing brain tumor segmentation method, the MRI preprocessing operation is introduced because it is directly connected to the quality of the segmentation results. In general, the original MRI picture must be preprocessed to achieve the purpose of segmentation. These preprocessing operations include noise reduction. Some preprocessed images are shown in Appendix A.4.

### 4.4 Feature Extraction

Some Feature Extraction images are show in appendix Fig.A.5.

#### 4.4.1 Mean (M)

The average value of picture is calculated in dividing all the pixel values of picture by the whole number of pixels in the picture classification, and then evaluating the treatment response. Listed below are several useful formulas of graphic features.

The average is just another name for the standard. To locate the average of a data position, add all the values and partition them by the number of values in the data set.

There are special types of averages, that is. Arithmetic mean, weighted average, geometric mean (GM) and harmonic mean (HM). If it is mentioned without an adjective (such as the mean), it normally refers to the arithmetic mean.

In probability and statistics, the general average or the expected value is a measure of the concentration tendency of a probability distribution or a random variable characterized by that distribution.

Eq.(4.1) 
$$M = \frac{1}{M \times N} \sum_{x=0}^{m-1} \sum_{y=0}^{n-1} f(x, y)$$

#### 4.4.2 Standard Deviation (SD)

The standard deviation is 2<sup>nd</sup> middle moment that describes the probability distribution of the observed people and can be used as a determine of homogeneity. advanced values indicate improved levels of concentration and high contrast at the edges of picture .

- In statistics, SD also represented by the lowercase Greek letter sigma  $\sigma$  of the population standard deviation or the Latin letter s of the standard deviation of the sample) is a calculate of the change or diffusion of a set of variables ideals.

- A low standard deviation indicate that values be inclined to be close to the set average IS called expected value), while a lofty standard deviation indicates that extend over a wider range.
- The standard deviation of a random variable, statistical population, data set or probability distribution is the square root of its variance.
- It simpler algebraically compared to the absolute mean deviation, but in reality it is less reliable.
- A helpfulland of the standard deviation is that, unlike variance, it is uttered in the same units as the data.

Eq. (4.2) 
$$M = \frac{1}{M \times N} \sum_{x=0}^{m-1} \sum_{y=0}^{n-1} f(x, y) |x, y|$$

#### 4.4.3 Homogeneity

The homogeneity of the variance is the underlying assumption of the t-test and the F-test (ANOVA), which takes into account the general variance (that is, the distribution or "distribution" of scores near the average) of two or more samples . It is equal to In correlation and regression, the term "uniformity of the variance of the matrix" (also known as homovariance) refers to the assumption that the variance of Y for each value of X is constant in the population. This entry focuses on homogeneity of variance as it relates to tests and ANOVA.

#### 4.4.4 Convolutional Neural Network

In machine learning, a convolutional neural network (CNN or ConvNet) is an artificial neural network that is often used for visual image analysis. In terms of their significance and the nature of their translation, they are also known as natural neural networks (SIANN) that are subject to migration or migration in space. The convolutional network is an inspiration in biological processes [4], because the relation of muscle tissue is similar to the muscle of the visual cortex of animals. A single cortical neuron responds to the stimulus in an infinite field, called the field of marriage. The various areas of neuron reception are closed, covering the entire field of vision. Compared to other image classification algorithms, CNN uses less preprocessing. This means that the handset will be able to censor the traditional network. The a priori knowledge and the independence of humans in the creation of features is the greatest advantage.

#### 4.5 Data Classification

In machine learning, Support Vector Machine is a learning model that monitors with an integrated learning algorithm that analyzes data used at Arizona State University to provide examples of training, each training instance is identified by one or both of these categories, the SVM training algorithm will develop a model that implements a new cononto method. The SVM model shows examples as points in space, and maps them so that the example shown in the symbol are given the same space by the new examples are expressed as categories. The edge of a cliff in the fall. In addition to line segmentation, SVM can also use the so-called kernel fish to record their access to high-speed sites, thus improving performance. Some graphical descriptions of data are shown in Appendix A.6.

#### 4.6 Performance Analysis

Some Performance Analysis images are show in appendix Fig.A.7.

- **True positive (TP)** = the number of cases correctly identified as patient.
- **False positive (FP)** = the number of cases incorrectly identified as patient.
- **True negative (TN)** = the number of cases correctly identified as healthy.
- **False negative (FN)** = the number of cases incorrectly identified as healthy.

**Accuracy:** The accuracy of a test is its ability to differentiate the patient and healthy cases correctly. To estimate the accuracy of a test, we should calculate the proportion of true positive and true negative in all evaluated cases. Mathematically

$$\text{Eq. (4.3)} \quad \text{Accuracy} = (\text{TP} + \text{TN}) / (\text{TP} + \text{TN} + \text{FP} + \text{FN})$$

**Sensitivity:** The sensitivity of a test is its ability to determine the patient cases correctly. To estimate it, we should calculate the proportion of true positive in patient cases. Mathematically, this can be stated as:

$$\text{Eq.(4.4)} \quad \text{Sensitivity} = (\text{TP}) / (\text{TP} + \text{FN})$$

**Specificity:** The specificity of a test is its ability to determine the healthy cases correctly. To estimate it, we should calculate the proportion of true negative in healthy cases. Mathematically, this can be stated as:

$$\text{Eq.(4.5)} \quad \text{Specificity} = (\text{TN}) / (\text{TN} + \text{FP})$$

#### 4.7 Summary

Medical images such as magnetic resonance imaging and computed tomography are an important way to effectively diagnose human diseases. The traditional method is to perform a manual analysis of brain tumors based on the visual inspection performed by a radiologist/doctor, which can lead to an incorrect classification when analyzing a large amount of MRI. To avoid human errors, a semi-automatic intelligent classification system is proposed. One of the main causes of human death are brain tumors. If the tumors are detected correctly at an early stage, the chances of survival can be increased. Magnetic Resonance Imaging (MRI) technology is used in the study of the human brain. In this process, the SVM is used as a classifier to detect AD.

Table 4.1 Result Performance

Dataset	Classifier	Accuracy	Sensitivity	Specificity
1	SVM	90%	100%	75%
2		94.59%	94.11%	100%
3		94.598%	94.21%	76%
4		95%	90.12%	100%
5		96%	93.11%	75%

To increase detection efficiency, different methods can be used for detection, then the best method is selected and parameters such as precision, sensitivity, specificity and accuracy are controlled. Experimental results show that the texture characteristics extracted by the wave atoms have a better classification rate. A classification technology based on the support vector machine (SVM) is proposed and applied to the classification of brain images. Therefore, the characteristics are extracted from the MRI images by means, standard deviations and homosexual convolutional neural networks. In this process, the maximum accuracy obtained by the SVM classifier is 90%, the specificity is 100% & the maximum is 100% .



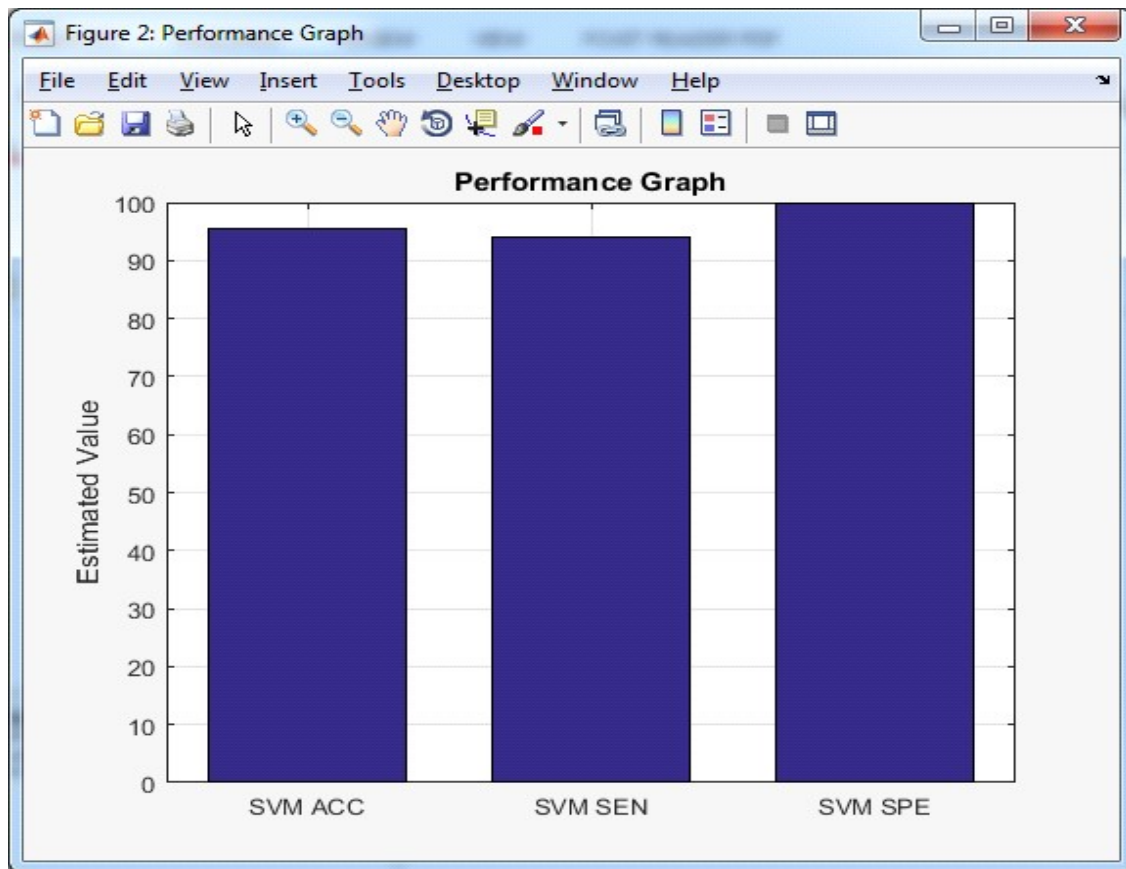


Fig. 4.2 Performance Graph

## **Chapter 5**

### **CONCLUSIONS AND FUTURE WORK**

One of the most challenging issues in Alzheimer's disease detection is the suspected correlation between the volume reductions of severity of the disease. This has motivated towards extracting texture features of the hippocampus for the task. In the present scenario, where the features such as CNN from whole brain image is used, this thesis brings forward the idea of using features such as wavelet coefficients, wave atoms coefficients along with CNN . The investigation demonstrates the improvement of classification accuracy as a result of the usage of wave atoms coefficients.

In order to evaluate the proposed approach for AD detection with texture features, data sets are selected and downloaded from ADNI. The proposed method is validated by measuring the parameters such as Sensitivity, Specificity and Accuracy.

SVM is used as a classifier to detect the AD. To improve the efficiency of detection, it is carried out using different methods and the best one is selected and monitoring the parameters such as Precision, Sensitivity, Specificity and Accuracy. The experimental results show that the texture features extracted using wave atoms give better classification rate.

#### **5.1 Conclusion**

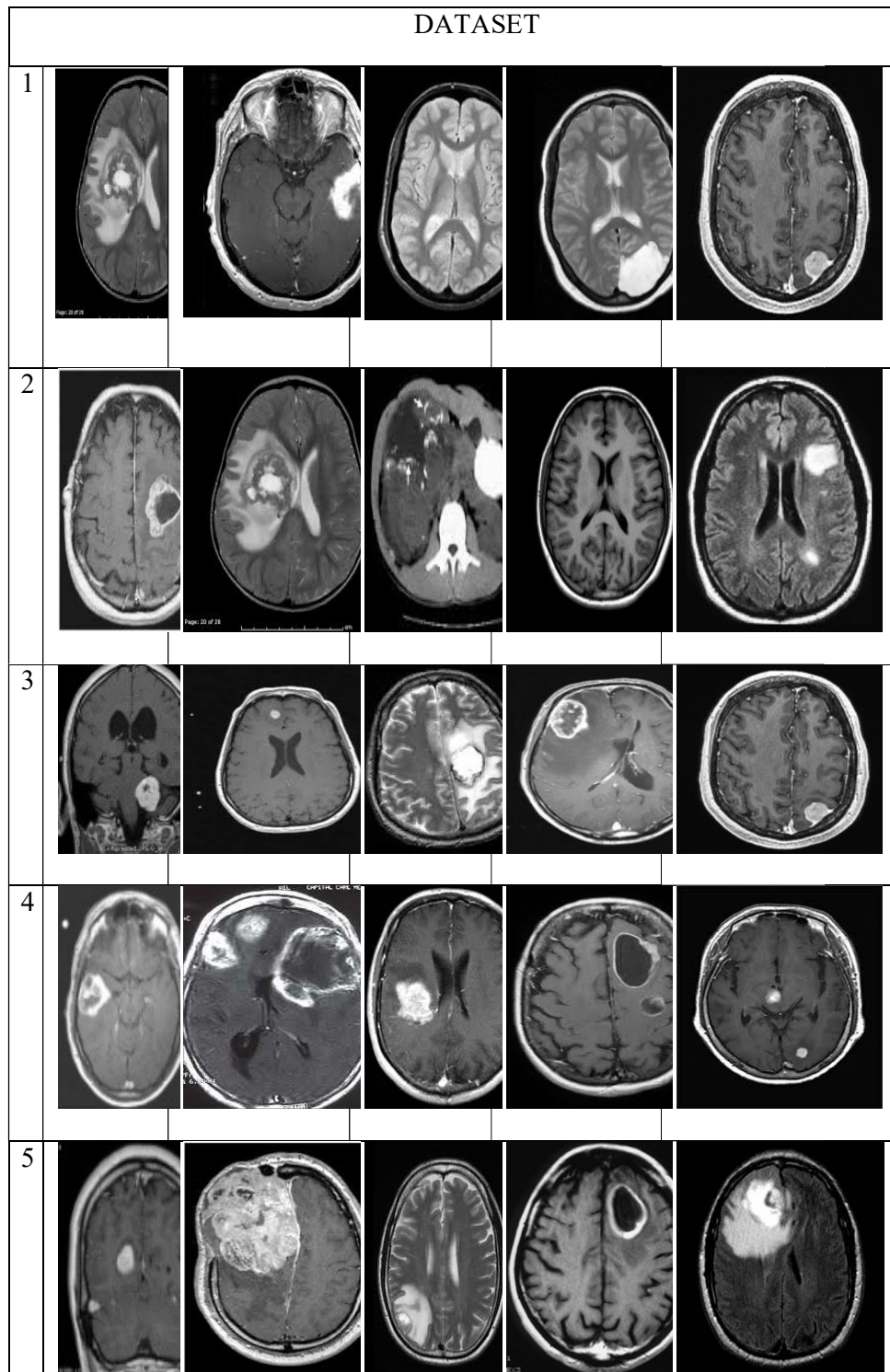
In case of 3D MRI from ADNI, research could not proceed to more Dementia classes because of lack of disease specific images above three types. Also could not add more number of samples, because every sample in 3D MRI sized around 40MB, thus significantly increasing processing burden. The task is rather simplified with use of 2D MRI from whole brain atlas database.

#### **5.2 Scope for Future Work**

Dementia classification can be taken towards more detail categories subject to the availability of Neuroimaging databases with sufficient number of cases of each and every disease type. Further work can be done to establish statistical significance through the use of numerous training sets as and when the databases get available with sufficient number samples. Also according to the opinion from Neurologist, results of this research if

combined with other clinical data, like onset of the disease or history it may lead to more robust diagnosis system.

## APPENDIX



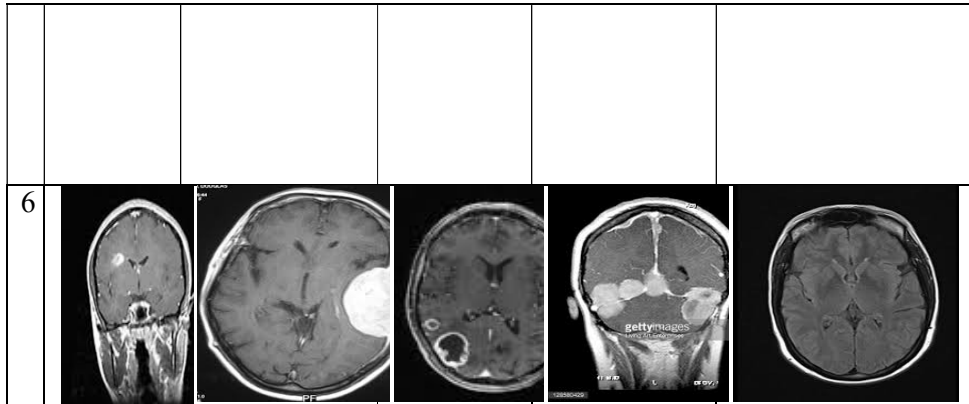


Figure A. 1 Data Set

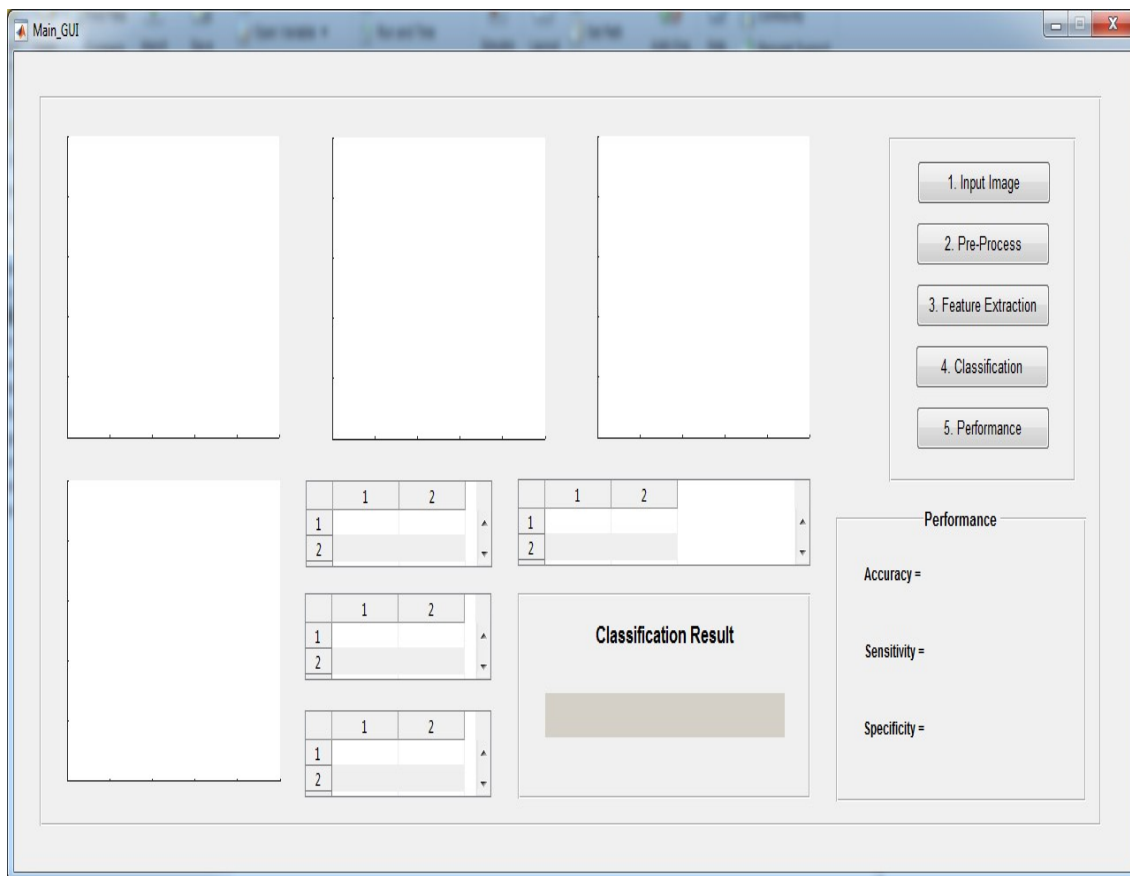


Fig. A. 2 Main Gui

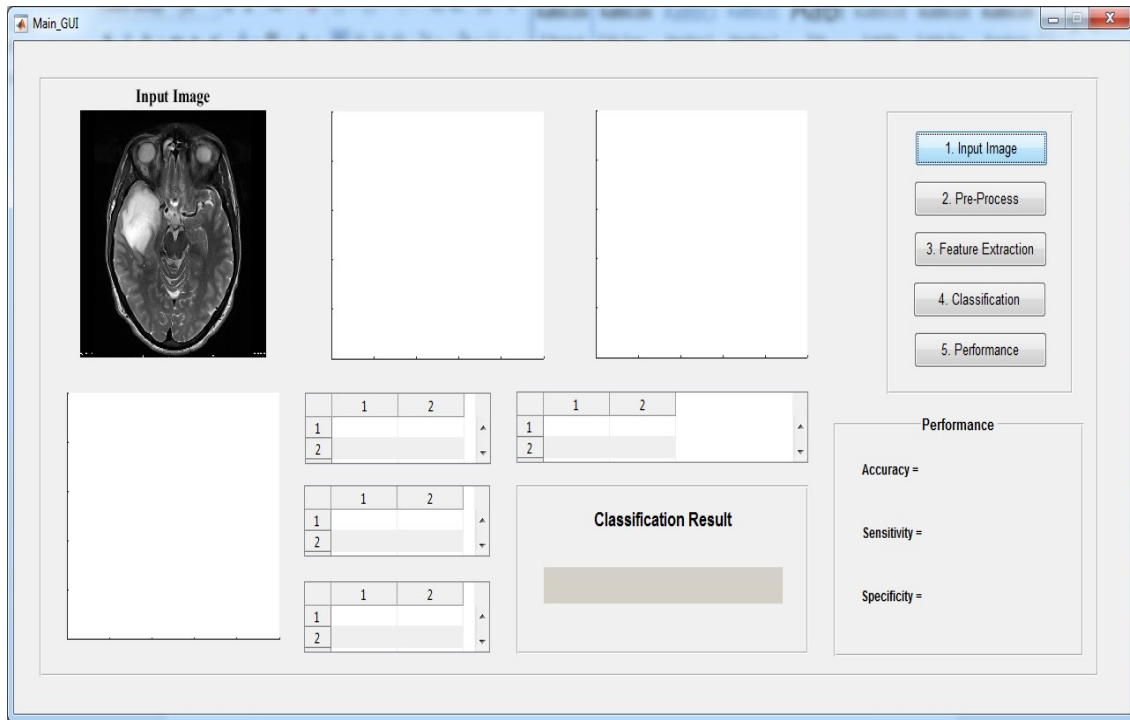


Figure A. 3 Input Image

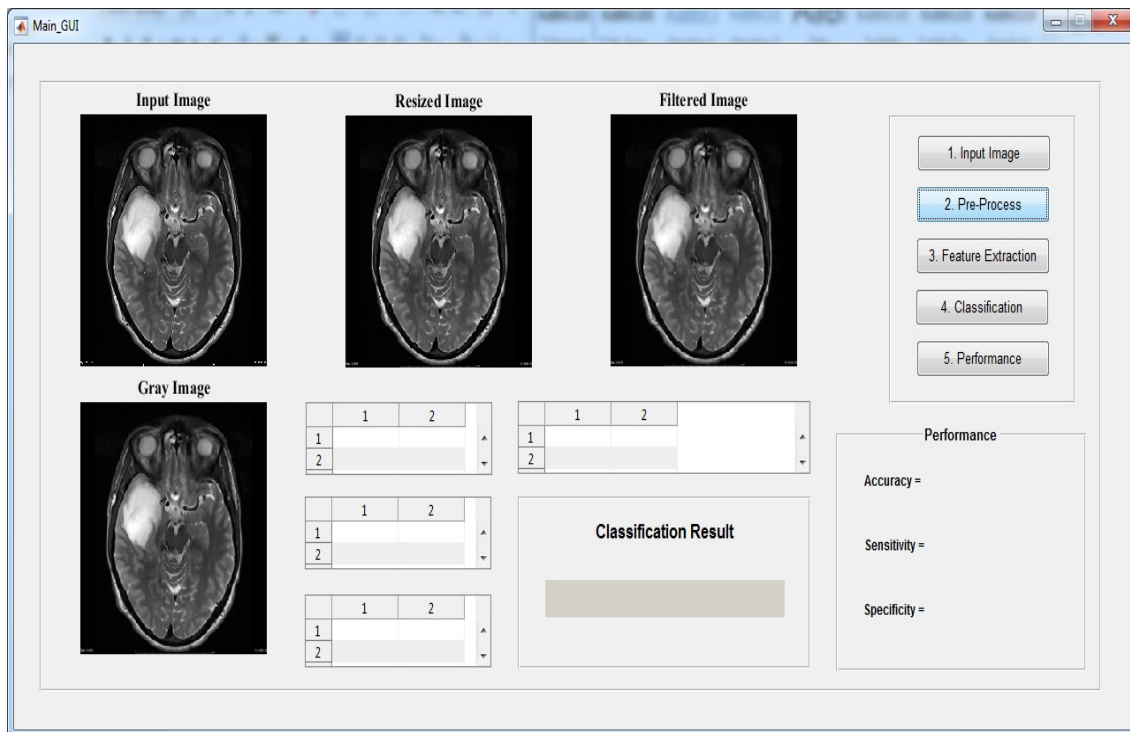


Figure A.4 Pre-Process Data

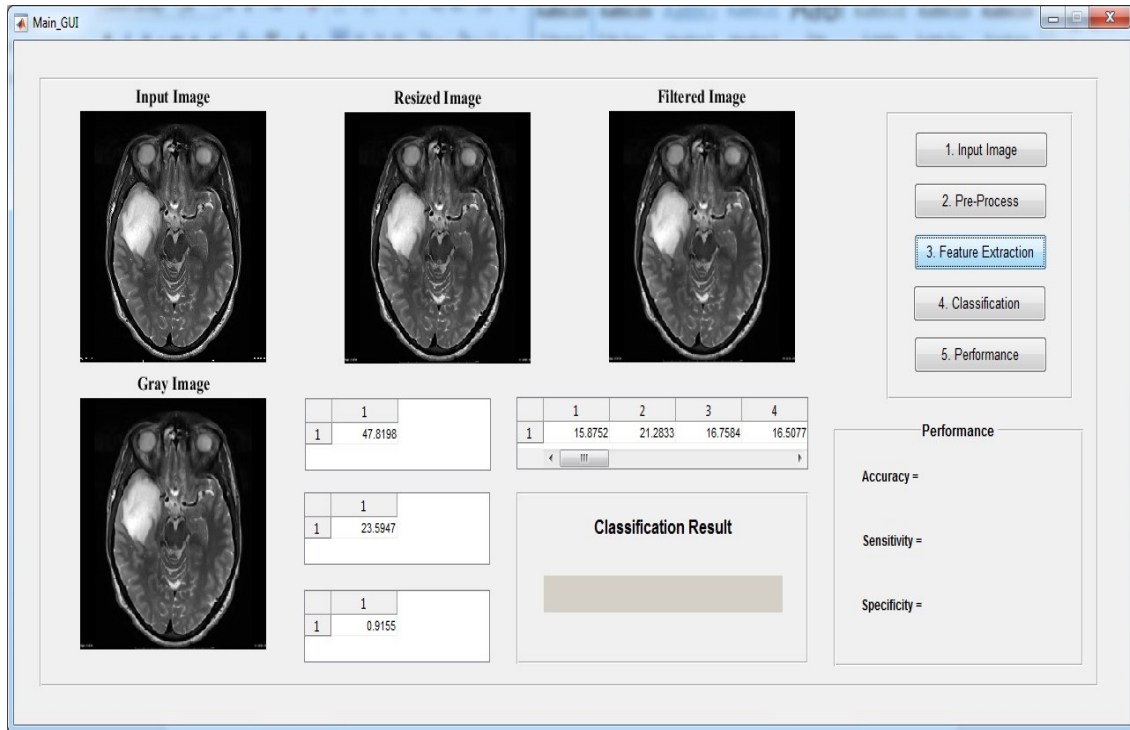


Figure A.5 Feature Extraction Data

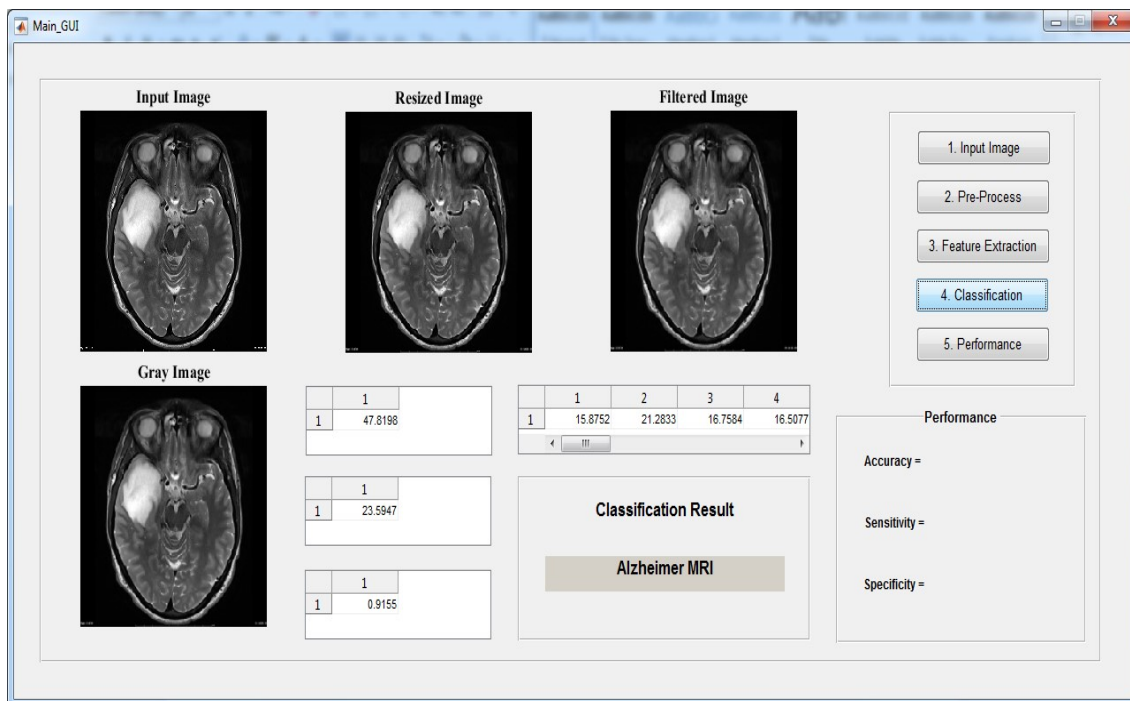


Figure A.6 Data Classification



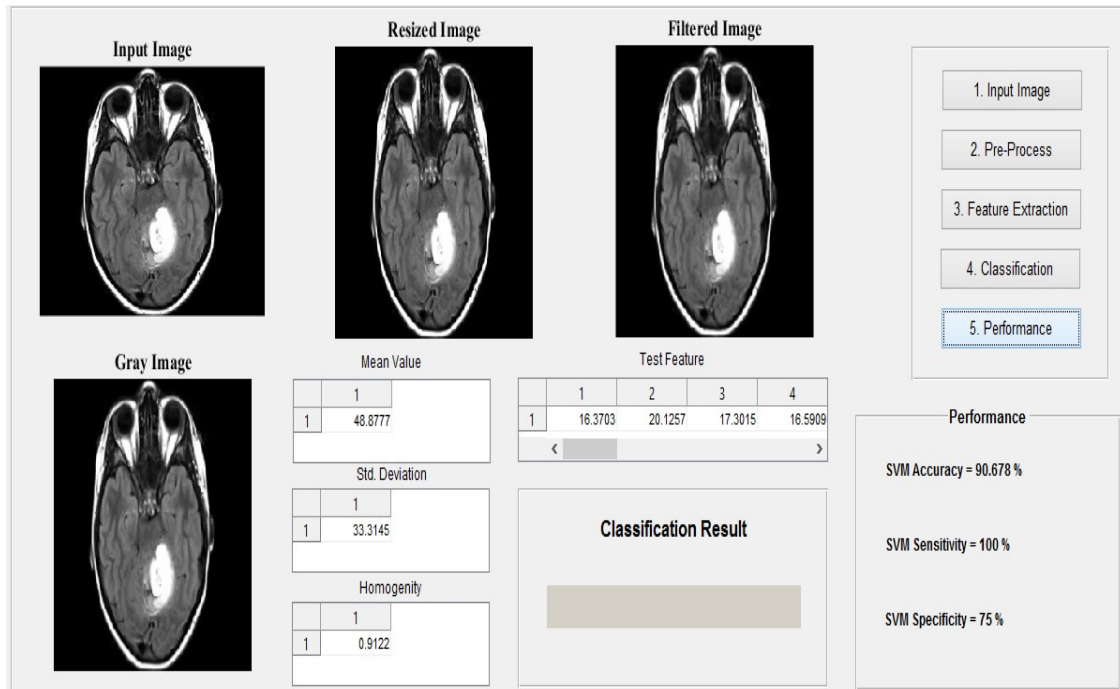


Figure A.7 Performance

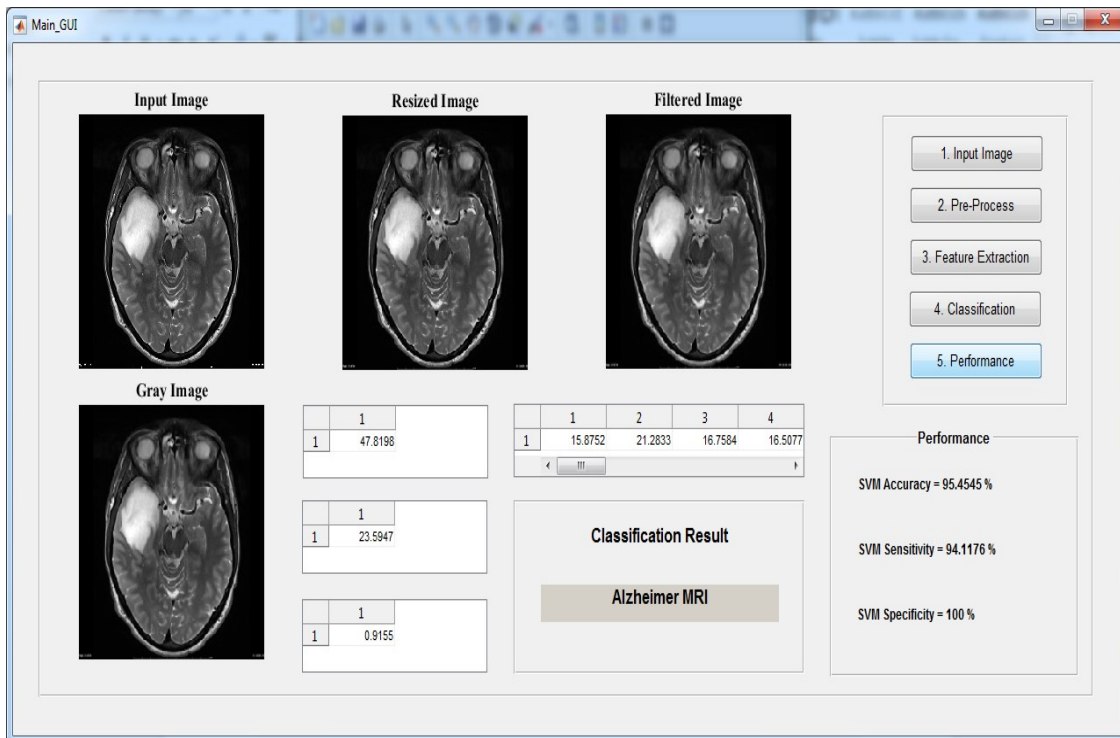


Figure A.8 Classification Result

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