



3D MR IMAGE PROCESSING USING MACHINE LEARNING TECHNIQUE FOR DETECTION OF ALZHEIMER'S DISEASE

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Abstract

Alzheimer's disease (AD) stands as a progressive neurodegenerative disorder, impacting millions worldwide and causing memory loss, cognitive decline, and compromised daily functioning. Early, precise diagnosis of AD plays a vital role in effective intervention and treatment planning. In recent years, deep learning methods, particularly Convolutional Neural Networks (CNNs), have exhibited promising outcomes in various medical image analysis tasks. In this research, we introduce an innovative approach for early Alzheimer's disease detection utilizing CNN-based deep learning models applied to neuroimaging data. Leveraging the potential of deep learning and CNNs, our method presents a dependable and non-invasive means of identifying AD in its early stages, facilitating timely medical interventions and improving the quality of life for affected individuals. Moreover, the proposed methodology can seamlessly integrate into existing healthcare frameworks, furnishing clinicians with valuable decision support tools for more precise and efficient Alzheimer's disease diagnosis.

Keywords:

Alzheimer's, classification, MRI, statistical features, vox-els, Machine Learning, Convolutional Neural Network.

I. Introduction

Alzheimer's disease (AD) is a debilitating neurodegenerative disorder characterized by progressive cognitive decline, memory loss, and impaired daily functioning. With the aging population globally, the prevalence of AD is increasing, posing significant challenges to healthcare systems and society as a whole. Early and accurate diagnosis of AD is crucial for timely intervention, patient care, and the development of effective treatment strategies. In recent years, the field of artificial intelligence (AI) and deep learning, in particular, has revolutionized various domains, including medical image analysis. Convolutional Neural Networks (CNNs), a class of deep learning models, have demonstrated exceptional capabilities in image recognition and pattern analysis. Leveraging the power of CNNs for the analysis of medical images offers a promising avenue for improving the accuracy and efficiency of AD diagnosis.

The study will explore the clinical implications of the developed CNN-based AD detection system. It will assess how the integration of this technology into clinical practice can enhance early diagnosis, aid in treatment planning, and improve patient outcomes.

This research focuses on harnessing the potential of CNNs for the early detection of Alzheimer's disease. By employing advanced deep learning techniques, this study aims to develop a robust and reliable AD detection system capable of accurately classifying patients with AD and distinguishing them from healthy individuals. The utilization of CNNs allows for the automatic extraction of intricate patterns and features from neuroimaging data, enabling precise and objective analysis.

This research represents a significant step toward leveraging cutting-edge technology to address the challenges associated with Alzheimer's disease diagnosis. By developing accurate and efficient CNN-based models, this study aims to contribute to the advancement of early detection methods, ultimately improving the lives of individuals affected by AD and easing the burden on healthcare systems worldwide.



II. Problem Definition

To global healthcare systems, with an increasing number of individuals affected by this neurodegenerative disorder. Early and accurate diagnosis of AD is critical for timely interventions, effective treatments, and improved patient outcomes. However, the current methods of AD diagnosis, particularly through medical imaging, are often labor-intensive, prone to inter-observer variability, and may not always enable early detection when it is most beneficial.

III. Literature Survey

Srinivasan Aruchamy, Amrita Haridasan, Ankit Verma, Partha Bhattacharjee, Sambhu Nath Nandy, and Siva Ram Krishna Vadali [1], in their research titled "Alzheimer's Disease Detection using Machine Learning Techniques in 3D MR Images," introduce a novel method for detecting Alzheimer's Disease (AD) by leveraging first-order statistical features in 3D brain Magnetic Resonance (MR) images. Alzheimer's disease is a neurodegenerative disorder primarily affecting elderly individuals, characterized by progressive cognitive decline. Early detection and classification of AD are crucial for effective disease management. Recent studies have employed voxel-based brain MR image feature extraction techniques coupled with machine learning algorithms to address this challenge. Given that both grey and white matter in the brain are impacted by AD pathology, analyzing both types of tissue proves to be more effective in disease prediction.

The proposed approach utilizes 3D structural brain MR images to differentiate between white and grey matter MR images. Subsequently, 2D slices are extracted in the coronal, sagittal, and axial planes, and key slices are selected for feature extraction. This methodology aims to enhance the accuracy and efficiency of AD detection by focusing on relevant structural features within the brain images.

Muhammad Hammad Memon, Jianping Li, Amin Ul Haq, and Muhammad Hunain Memon [2], in their study titled "Early Stage Alzheimer's Disease Diagnosis Method," introduce a machine learning-based approach for accurately diagnosing Alzheimer's disease in its early stages. The researchers employed machine learning classifiers to predict Alzheimer's disease with a high degree of accuracy. To assess the effectiveness of their proposed method, they utilized the Alzheimer's Disease Neuroimaging Initiative dataset. This dataset was instrumental in evaluating the performance of the proposed method in accurately diagnosing Alzheimer's disease at an early stage.

Chima S. Eke, Emmanuel Jammeh, Xinzhong Li, Camille Carroll, Stephen Pearson, and Emmanuel Ifeakor [3], in their study titled "Early Detection of Alzheimer's Disease with Blood Plasma Proteins Using Support Vector Machines," propose a methodology aimed at identifying potential non-amyloid biomarkers in blood for the early detection of Alzheimer's disease (AD). The authors highlight the appeal of using blood-based biomarkers due to their accessibility and cost-effectiveness. Their approach primarily revolves around the utilization of machine learning (ML) techniques, with a specific emphasis on support vector machines (SVMs). SVMs are chosen for their capability to construct multivariable models by discerning patterns from intricate datasets.

S. Kalpana Devi, D. Amirthavarshini, RS Anbukani, and S. Bhavatha Ranjanni [4], in their project titled "Personal Assistance for Alzheimer's Patient," start by delineating Alzheimer's disease as a progressive neurological disorder characterized by the gradual deterioration of memory and cognitive functions. They note that contemporary approaches to managing this disease involve platforms for discussion, behavioral monitoring, and path tracking.

Their proposed solution entails developing a mobile application that serves as a personal assistant for Alzheimer's patients. The application's features encompass facial recognition, detection of wandering and fainting episodes, assistance in navigating back home, reminders for daily tasks and past events, as well as organizing and planning activities. They suggest leveraging smartphone sensors and GPS capabilities to monitor patient actions. The implementation involves utilizing deep learning and machine learning techniques using Python within the Android Studio environment, integrated with



Google Maps functionalities. Additionally, the inclusion of specially designed games within the application aims to provide cognitive stimulation and support during the early stages of the disease..

Hala Ahmed, Hassan Soliman, and Mohammed Elmogy [5], In their study titled "Early Detection of Alzheimer's Disease Based on Single Nucleotide Polymorphisms (SNPs) Analysis and Machine Learning Techniques," underscore Alzheimer's disease (AD) as a debilitating neurodegenerative condition that worsens over time. They emphasize the significance of genetic factors, particularly the Apolipoprotein E (APOE) gene, which has been consistently associated with AD in various genome-wide association studies (GWAS). Single nucleotide polymorphisms (SNPs), the most common type of genetic variation among individuals, are recognized as significant biomarkers for AD. Leveraging SNPs in the early stages of the disease aids in understanding and detecting AD.

The primary objective of their paper is to achieve early prediction and diagnosis of AD with high classification accuracy by identifying SNPs biomarkers associated with the disease. The study focuses on utilizing machine learning (ML) techniques to identify these AD biomarkers. Naïve Bayes (NB), Random Forest (RF), Logistic Regression (LR), and Support Vector Machine (SVM) algorithms are applied to AD genetic data from the Alzheimer's Disease Neuroimaging Initiative phase 1 (ADNI-1) and Whole-genome sequencing (WGS) datasets.

Aakash Shah, Dhruvi Lalakiya, Shekha Desai, shreya and Vibha Patel [6], In their study titled "Early Detection of Alzheimer's Disease Using Various Machine Learning Techniques: A Comparative Study," start by delineating Alzheimer's disease (AD) is a progressive neurodegenerative disorder characterized by memory loss and cognitive decline. While various psychological tests are employed by healthcare professionals to diagnose AD, there is currently no universally accepted method for detecting the disease in its early stages. This paper presents a comprehensive analysis of the efficacy and accuracy of different machine learning techniques applied to a combination of biomarkers associated with AD.

One novel approach explored in this study is the utilization of the Voting Classifier Algorithm for early detection of AD. Unlike previous research, which has not extensively examined this algorithm for AD detection, the present study aims to elucidate the impact of the averaging factor inherent in the Soft Voting Classifier. This factor influences the accuracy of the algorithm's output by aggregating the predicted outcomes of multiple classifiers, thereby reducing the potential for inaccuracies and producing more precise results.

By employing the Soft Voting Classifier, this study seeks to enhance the reliability and accuracy of AD detection by leveraging the collective wisdom of diverse classifiers. Through this approach, the algorithm combines the strengths of individual classifiers, mitigating the weaknesses inherent in any single model and yielding a more robust diagnostic tool for identifying AD in its early stages.

IV. Existing System

Alzheimer's Disease (AD) poses a significant global health challenge, necessitating early and accurate detection for effective intervention. Leveraging advancements in machine learning and medical imaging, researchers have developed sophisticated systems aimed at detecting AD through the analysis of 3D Magnetic Resonance Imaging (MRI) data. This approach offers a non-invasive means to identify subtle structural changes in the brain associated with the disease progression. The existing system employs a multifaceted approach that integrates state-of-the-art machine learning algorithms with advanced image processing techniques tailored for 3D MRI analysis.

V. Drawbacks

One significant challenge is the availability of comprehensive and diverse datasets. Many machine learning models used for Alzheimer's detection operate as black boxes, making it challenging to interpret their decisions. Despite model validation, over fitting remains a concern.

VI. Proposed System

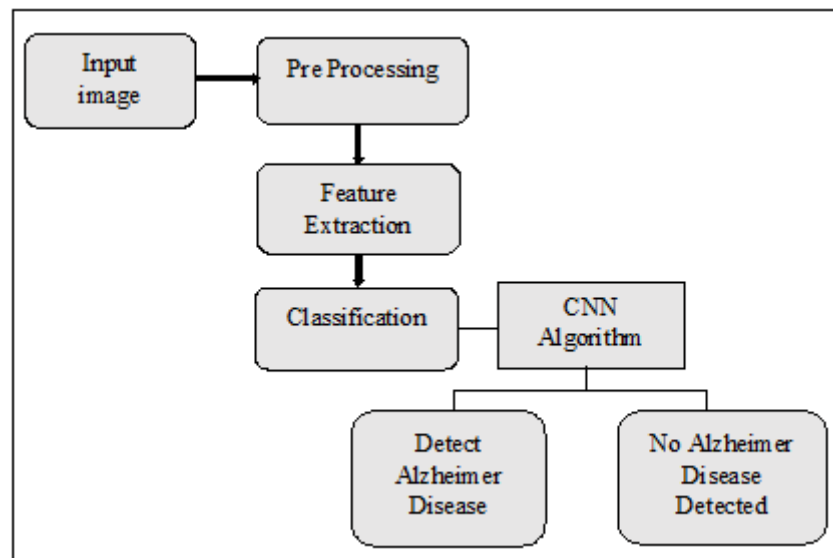


Figure 1: Proposed system

6.1. Input: The first stage involves gathering data, which is critical since the quality and quantity of information collected directly impact the scope and accuracy of your prediction model.

6.2. Data pre-processing: Data pre-processing includes visualizing information to identify relationships among parameters in the dataset and addressing data imbalances. Furthermore, data must be divided into two parts: one for training the model and the other for testing.

6.3. Feature Extraction: In the next phase of our workflow, we focus on feature selection. Various models have been utilized by researchers and scientists for different purposes, such as image processing and sequence analysis (like text, numbers, or patterns). In our case, we've gathered samples from individuals with Alzheimer's disease. As a result, we choose models that can differentiate between patients with Alzheimer's and healthy individuals.

6.4. CNN model: Convolutional Neural Networks (CNNs) excel in handling grid-like data, making them ideal for image analysis tasks. Their ability to automatically detect patterns relevant to Alzheimer's disease within image data is notable. Additionally, CNNs have the potential to identify subtle movement patterns associated with Alzheimer's disease through thorough analysis.

6.5. Prediction: In this phase, we prepare the model to accurately detect the presence of Alzheimer's disease using the provided dataset.

VII. Technology

Python is widely utilized across various domains including server-side web development, software engineering, mathematical computations, and system scripting. Its popularity stems from its suitability for Rapid Application Development (RAD) and its role as a scripting or integration language for connecting existing components. Python's high-level nature, comprehensive built-in data structures, dynamic typing, and dynamic binding contribute to its appeal. The language's readability and straightforward syntax are instrumental in reducing program maintenance costs. Moreover, Python's support for modules and packages encourages modular programming and facilitates code reusability. As an open-source language with a vibrant community, Python continuously benefits from contributions by independent programmers who develop libraries and extend its functionalities. In professional settings, Python is favored for backend web development, data analysis, artificial intelligence, and scientific computing. Additionally, developers leverage Python to create productivity tools, games, and desktop applications.

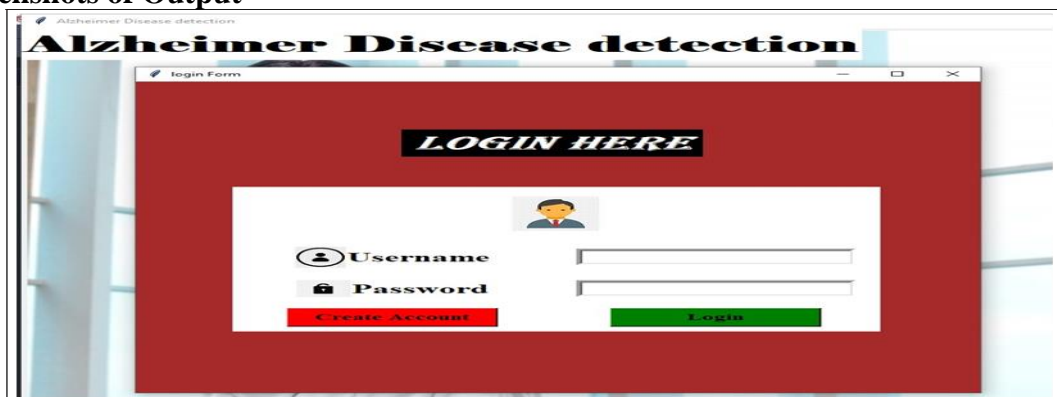
VIII. Algorithm

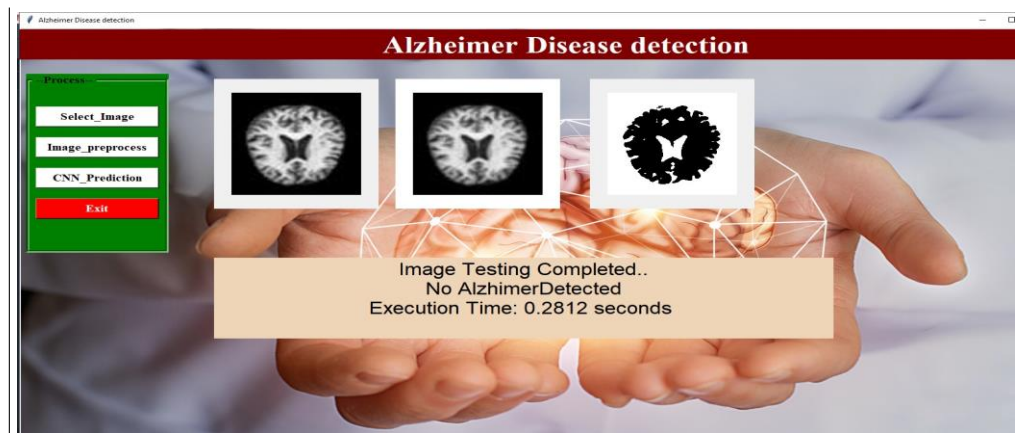
A Convolutional Neural Network (CNN) is a popular architecture in Deep Learning, commonly applied in Computer Vision, a branch of Artificial Intelligence focused on enabling computers to understand and interpret visual data, particularly images. CNNs excel at automatically learning hierarchical representations from images. •The 3D nature of MRI scans contains spatial information across multiple dimensions. CNNs are naturally suited to handle spatial hierarchies present in 3D data, allowing them to learn spatial relationships between voxels (3D pixels) within the brain volume.

In a standard Neural Network, there are three primary types of layers:

1. **Input Layers:** These layers serve as the entry point for data into the model. The number of neurons in this layer corresponds to the total number of features in the dataset. For images, this typically aligns with the number of pixels.
2. **Hidden Layer** Following the Input layer, data is passed to the hidden layers. The number of hidden layers and neurons within each layer can vary depending on the model and dataset dimensions. Each hidden layer applies matrix multiplication to the output of the previous layer using specific learnable weights, followed by the addition of learnable biases. Subsequently, an activation function is applied to introduce nonlinearity to the network.
3. **Output Layer:** The output from the hidden layers undergoes further processing through a logistic function, such as sigmoid or softmax. This transformation converts the output of each class into a probability score, facilitating classification.

IX .Screenshots of Output





X. Conclusion

The research on Alzheimer's Disease Detection using Convolutional Neural Networks (CNN) represents a significant advancement in the field of medical image analysis and neurodegenerative disease diagnosis. Through a comprehensive methodology that involved data collection, pre-processing, model development, training, and evaluation, this study aimed to leverage the power of CNNs to provide accurate, interpretable, and early detection of Alzheimer's disease. The research on Alzheimer's Disease Detection using CNN not only pushed the boundaries of technology but also emphasized the importance of ethical and interpretable AI in healthcare. By providing clinicians with a powerful, reliable, and transparent tool for early Alzheimer's disease diagnosis, this study has opened new avenues for improving patient outcomes and advancing the field of neurodegenerative disease research. As we move forward, the lessons learned from this research will continue to shape the future



of AI-driven healthcare, leading to more accurate diagnoses, personalized treatments, and ultimately, better lives for individuals affected by Alzheimer's disease and other neurodegenerative disorders.

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