

Regression and Multi-Class Classification of Alzheimer's Disease Diagnosis Using NMF-TDNet Features From 3D Brain MR Image

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Abstract—To more accurately depict Alzheimer's disease (AD) and predict clinical scores while taking into account advancements in clinical imaging and substantial learning, several experts are gradually using ConvNet (CNNs) to remove deep level features from clinical images. A small deep learning network called the PCANet (Principal Component Analysis Network) creates multi-faceted channel banks to verify the accuracy of voluminous head part assessments. After binarization, blockwise histograms are constructed to obtain picture properties. PCANet is less adaptable because multi-facet channel banks are built with test data, resulting in PCANet features with thousands or even millions of aspects. The NMF-TDNet (non-negative matrix factorization tensor decomposition network) is an information-free organization based on PCANet that we present in this study to address these issues. Instead of PCA, staggered channel banks are made to test nonnegative matrix factorization (NMF). By applying tensor decomposition (TD) to a higher-demand tensor derived from the learning results, the input's dimensionality is reduced, resulting in the final image features. The support vector machine (SVM) in our technique uses these properties as input to diagnose, predict clinical score, and categorise AD.

Keywords—There are many different names for Alzheimer's disease (AD), such as deep-learning, regression, and multi-class classification.

I. INTRODUCTION

Alzheimer's is a neurodegenerative state with a protracted course that often affects the elderly. As the infection advances, the patient's memory and mental capabilities decay, their neurons are bit by bit obliterated, and the patient in the long run passes on [1]. Alzheimer's disease influences around 50 million individuals around the world. As the total populace ages, it is guessed that the quantity of Alzheimer's victims will twofold by 2050 [2, 3]. While there are a few prescriptions available to treat Alzheimer's disease, their viability is restricted to easing back the infection's movement as opposed to relieving it out and out [4]. The patient's early mental impairment falls somewhere amidst CN i.e., cognitive normal state and the Alzheimer's illness condition, also known as mild cognitive impairment state (MCI), according to various evaluations. Diverse researches are actively going on to find people who have proactively gone to the MCI stage in order to provide the appropriate mediation to halt the condition's progression [5]. Thus, deciding the phase of Alzheimer's disease has been the essential focal point of flow research, and early location of the infection is significant.

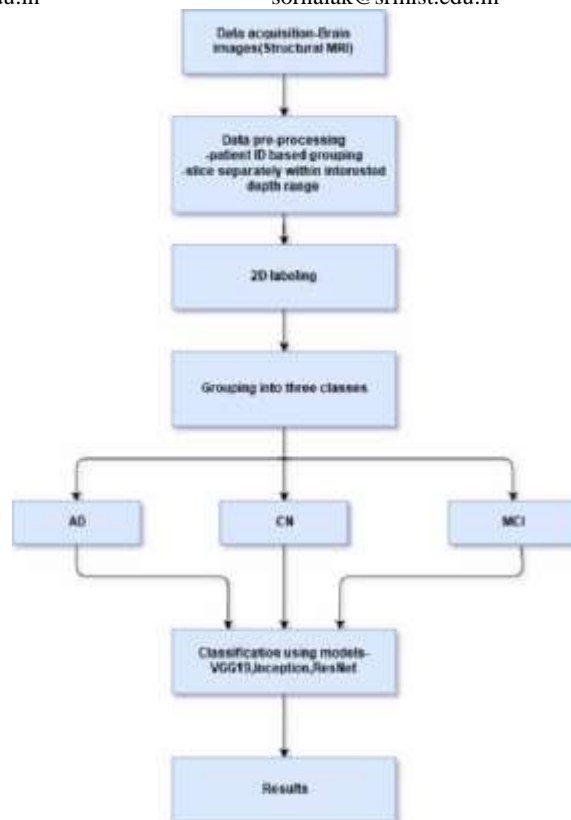


Fig.1: Example figure

Medical imaging technology has advanced tremendously in recent years. It will likely give clinical experts and scholastics with an interesting perspective on disorder finding by means of the investigation of clinical samples, confirming the legitimacy of clinical research expert's conclusions, and giving extra data to advance further review and examination. Diverse clinical image methods (some of the widely known methods are positron emission tomography (PET), single-photon emission computed tomography (SPECT), and magnetic resonance imaging (MRI)) can be used to obtain clinical images. The quick evaluation of changes, a top concern in construction and processing, as well as the identification of biomarkers for Alzheimer's disease may benefit from these numerous image techniques.

A few tests have shown that MRI is one of the most common and well-known imaging modalities used in medical practice. Neuron incident is the purest indication of

Alzheimer's disease pathology, trailed by mind shrinkage from AD explicit frontal cortex districts to the whole cortical locale. An MRI may be used to observe these changes. These conspicuous physical changes happen preceding the beginning of huge mental limit decline. Subsequently, most of flow research is centered around laying out a patient's Promotion stage utilizing X-ray based PC supported conclusion (X-ray computer aided design).

II. LITERATURE REVIEW

Nowadays most of the people are busy with their fast-paced lives. Which in turn is the reason for lots of people suffering with high blood pressure making it as the root cause of dementia cases increment. Existing Alzheimer's Disease medication can cause a lot of side effects to the patients like vomiting sensation, nausea, diarrhea, allergic reactions on body, loss of appetite, headaches, dizziness and confusion. Regardless of lots of research and development the fundamental systems of Alzheimer's infection and applicable treatment targets stay muddled, and there is no proper drug which produces low or few side effects is available. There are as of now research focusing on amyloid and tau, two principal pathology signs of Alzheimer's disease. Yet, additional studies exist in various stages of research that are largely focused on other signs and processes connected with the condition that are now being inspected. Issues covered include: The study by **A. Khan, and C. Ballard, et al.**, looks at novel and drug repositioning that are now being explored in clinical trials that spot non-tau and non-amyloid pathways. This includes both possible disease-modifying medications and cognitive and neuropsychiatric therapies. The publications included in this study were obtained using PubMed and clinical trials database searches. Researchers are working harder than ever to find effective treatment for the behavioral signs of Alzheimer's disease, employing both new medications and drug repositioning. Other routes are being investigated as a consequence of the failure of medications. Upgrades in biomarkers will give additional scope for clinical preliminaries of promising drugs for meaningful treatment and disease adjustment in Alzheimer's disease.

K. G. Yiannopoulou, and S. G. Papageorgiou., discuss that, the groundwork of contemporary Alzheimer's disease (AD) therapy shows restraint focused psychoeducation, shared objective setting, and independent direction considering serious core competencies for a connection between the doctor, patient, and the care taker. Utilized as part of a comprehensive treatment plan, cholinesterase inhibitors (ChEIs), also known as N-methyl-d-aspartate (NMDA) antagonists, and other FDA-supported promotion medications like meantime may have limited "disease course-modifying" advantages by enhancing perception and reducing loss of freedom. Joining pharmacologic and nonpharmacologic treatment might diminish side effects essentially, forestall clinical movement, and decrease absolute medical services costs. The most important phase in Alzheimer's disease pharmacotherapy is to distinguish and eliminate any potentially hazardous prescriptions and enhancements. Nonpharmacological treatments are utilized as the principal

method of medication for neuropsychiatric secondary responses and risky process of behaving. Among the techniques utilized incorporate psycho education, trigger location, iterative assessment, and conduct and ecological mediation changes. Significant exploration is being performed to deliver better medications, restorative devices, and more exact and accommodating indicative biomarkers for Alzheimer's disease. Various remedial targets, including neurotransmitter, tau neurotic-cycle and amyloid, power house of cell, provocative pathways, neuroglia, and multiple model ways of life mediations, are the focal point of continuous exploration reads up for Alzheimer's disease essential and auxiliary anticipation, in addition to clinical preliminaries assessing suggestive and sickness adjusting therapies in Alzheimer's disease patients.

In their work, **T. Tong, R. Wolz, Q. Gao, R. Guerrero., et al.**, survey the machine learning strategies have been broadly used to aid the assessment of neurological circumstances like dementia and to recognize morphological anomalies in essential frontal cortex appealing resonance imaging information. In this paper, we present a program for the beginning phases of Alzheimer's sickness and gentle mental impedance that joins a numerous occasion learning (MIL) technique (MCI). Neighborhood power patches are disposed of as highlights in our examination. Nonetheless, not each of the patches given out by dementia victims are gone after in much the same way, and some may not display the sickness' normal symptoms. As an outcome, pinpointing these patches as transporters of specific disorders might be troublesome. Managed learning frameworks, like MIL, might have the option to manage the issue of equivocal preparation marks. A diagram is created for each picture to take advantage of the associations between the patches and consequently, tackle the MIL issue. The created diagrams remember data for how patches show up and associate with each other, which might mirror the fundamental designs of the pictures and aid arrangement. The proposed method was able to accurately depict 89% of AD victims and sound controls and 70% of patients with stable MCI and mild MCI using benchmark MR evaluations from 834 ADNI centers. A well-informed perspective on the diagnosis and treatment of neurodegenerative diseases could result from the proposed method's ability to produce outcomes that are comparable to or superior to those of two cutting-edge methods that make use of the same dataset.

In a recent study on the diagnosis of Alzheimer's disease and MCI **D. Shen, H. Suk, and X. Zhu et al.**, a significant correlation was found between the diagnosis and the clinical score assumption. Additionally, it has been demonstrated that the issues of low model size and high component dimensionality can be resolved by incorporating decision-making through complex learning model. Clinical score regression and clinical imprint collection were frequently restricted prior to evaluations alone. Supposedly, most of past component determination research considers unfortunate capacity, that I termed as the contrast between the objective and anticipated values, component by component. In this paper, we take a gander at the problem with changing the joint and grouping for the finding of AD/MCI utilizing an original grid similitude-based

misfortune capability that utilizes the significant data level in the objective reaction network and requires data to be protected in the anticipated reaction lattice. The recently developed capability is used to relate the collection tether way to deal with pick joint highlights across undertakings, for example, clinical score expectation and class marking. The suggested method was tested using the Alzheimer's Neuroimaging Initiative (ADNI) dataset. The results showed that the really transmitted catastrophe limit performed better than the cutting-edge methods in clinical score suspicion and confusion status perceiving proof.

Aim of the Study by **L. Hou, X. Zhang, K. Hu, Y. Wang, K. Chen, et al.**, is to find out the patients suffering with moderate cognitive impairment (MCI) who have been promoted to Alzheimer's disease (AD) and those who have not seen a regular doctor in three years by utilizing multi-scale credits obtained from benchmark essential attractive resonance imaging. 228 normal controls (NC) and 133 MCI patients are included, 71 of whom switched to advancement within three years or less. General representation of 549 patients suffering from the Alzheimer's are collected as Neuroimagery Initiative (ADNI) data set, that individuals are termed as to MCI converters or MCIc, and 188 Advancement patients are recalled for the purpose of the examination. Using the standard voxel-based morphometry strategy, the pictures are preprocessed by separating the cerebrospinal material, white and dim matter. The wavelet outline, a multi-scale image portrayal technique, is utilized to separate credits of different sizes and directions from the handled dim matter picture. Both whole dark matter imaging and hippocampal dim matter pictures might be utilized to create highlights. The support vector machine (MCInc) is utilized in the creation of classifiers for MCIc and MCI non-converters. Utilizing nearby hippocampus information, the precision for arranging Promotion versus NC and MCIc versus MCInc utilizing a leave-one-out method is 84.13 percent and 76.69 percent, individually. Research discoveries show that the suggested multi-scale methodology is fit for recognizing MCI converters and non-converters, and that it has a decent possibility being viable for MCI assumption in clinical environment.

In light of non-negative matrix factorization (NMF) evaluation (AD), this letter introduces yet another computer-based diagnostic (CAD) strategy for the early detection of Alzheimer's disease prediction by using single photon emission computed tomography (SPECT) images **P. Padilla et al.**, This exploration utilized a SPECT information base that included standardized information from selected patients as well as sound reference individuals. The SPECT informational index is assessed using Fisher discriminant ratio (FDR) and the feature extraction NMF for each subject's applicable parts. These preprocessing techniques are fundamentally planned to bring down the serious level of aspect in the upcoming information and mitigate the supposed "revile of dimensionality" issue. The NMF-transformed set of data, which has fewer highlights, is grouped using an SVM-based characterization strategy. The NMF combined with SVM approach accurately perceives SPECT pictures with up to 94% exactness and high responsiveness and explicitness values (over 90%). For

fulfillment, an examination between the proposed technique and another as of late evolved PCA in addition to SVM strategy is given. The outcomes show that the NMF+SVM procedure beats the PCA+SVM and customary voxel-as-feature (VAF) as well as SVM methods.

III. METHODOLOGY

As deep learning and clinical imagery advancements have progressed, few researchers are currently using convolutional neural networks (CNNs) to remove deep level qualities from clinical photos to all the more precisely order clinical scores and Alzheimer's disease (AD). A small deep learning network called the PCANet uses principal component analysis (PCA) to create multi-layer filter channel banks for testing. To get picture properties, blockwise histograms are developed after binarization. PCANet is less flexible since the multi-facet channel banks are constructed utilizing information from testing, bringing about PCANet highlights with several thousands or even countless aspects.

Disadvantage

1. Sample data is required for the creation of multilayer filter banks.
2. Restricting the flexibility of PCANet.

In this research, we come up with the proposal of data-free PCANet-based nonnegative(excluding positive and zero) matrix factorization tensor decay organization, or NMF-TDNet, as a solution to these issues. Instead of PCA, staggered channel banks are made to test with NMF. Using tensor decomposition (TD) to reduce the data's dimensionality and fabricating a tensor with a larger demand, the final image characteristics are produced. All in all, we utilize these properties as a commitment to the support vector machine (SVM) to decide advancement arranges and expecting clinical scores. The Desert spring, ADNI-1, and ADNI-2 datasets are utilized to assess our technique's presentation fittingly.

Advantages

1. The experimental findings suggest that NMF-TDNet has the potential to reduce data dimensionality.
2. When NMFTDNet features were used as input, the results were better than when PCANet features were used as input.



Fig.2: System architecture diagram

Modules

To finish this project, we utilized the following modules:

- Data exploration: we will use this module to import the data into the working system.
- This module then used as a tool to read data for processing.
- Train Test splitting: Data is segregated into train and test data using this python module..
- The models were created using Mobilenet, InceptionResnetv2, SVM embedded with ConvNet layer, SVM classifier, and support vector regression. Algorithm accuracy was calculated.
- User registration and login: Accessing this module requires registration and login.
- The use of this module will result in anticipated input.
- Prediction: the final predicted value is shown.

IV. IMPLEMENTATION

4.1 Algorithms

Mobilenet

Mobile Net and other convolutional neural networks have been designed specifically for use in embedded and mobile vision applications. They depend on a proficient engineering that forms conservative profound brain networks with insignificant idleness for installed and cell phones utilizing depthwise detachable convolutions.

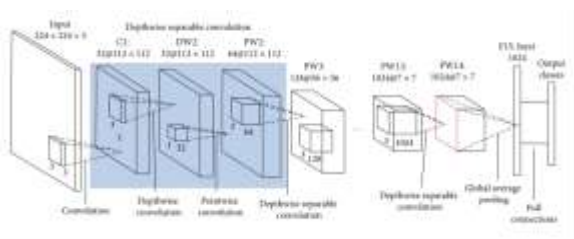


Fig.3: Mobilenet architecture

InceptionResnetv2

More than a million images from the ImageNet collection were utilised to train the Inception-ResNet-v2 convolutional neural network. The 164-layer organization can sort pictures into 1000 different item classifications like console, mouse, pencil, and various creatures.

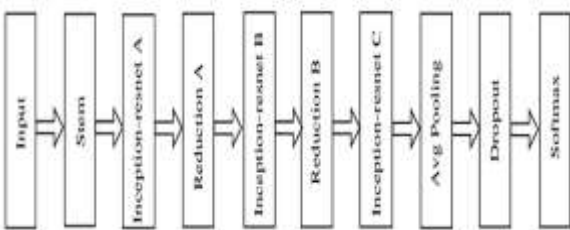


Fig.4: InceptionResnetV2 model Diagram

SVM embedded with CNN layer

Support vector machines (SVM) and convolutional neural networks (CNN) for picture order infrastructure. Fred

Abien, Agarap. Convolutional neural networks (CNNs), like "normal" brain organizations, are made out of neural network layers of neurons with "learnable" boundaries.

Support vector classifier

A deep learning framework known as a support vector machine (SVM) makes use of managed learning to recognize or anticipate the behavior of a data set. Artificial Intelligence and Machine Learning with supervised learning systems give input and desired output for categorization.

Support vector regression

A managed learning approach called as help vector relapse is utilized to foresee discrete qualities. A similar hypothesis is utilized by SVMs and Support Vector Regression. SVR's significant objective is to recognize the ideal fit line. The hyperplane with the most focuses is the best fit line in SVR.

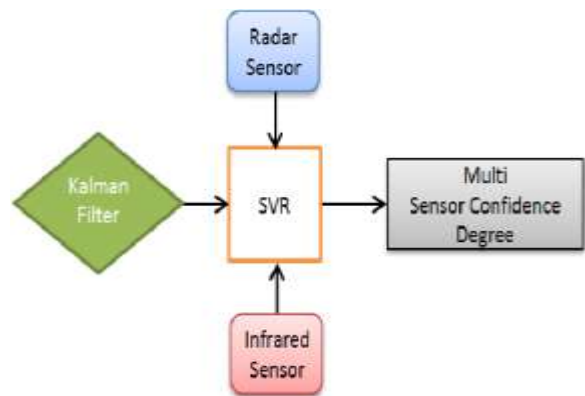


Fig.5: SVR model

V. EXPERIMENTAL RESULTS



Fig.6: Home screen View



Fig.7: User signup page View



Fig.8: User sign-in page View



Fig.9: Main page View

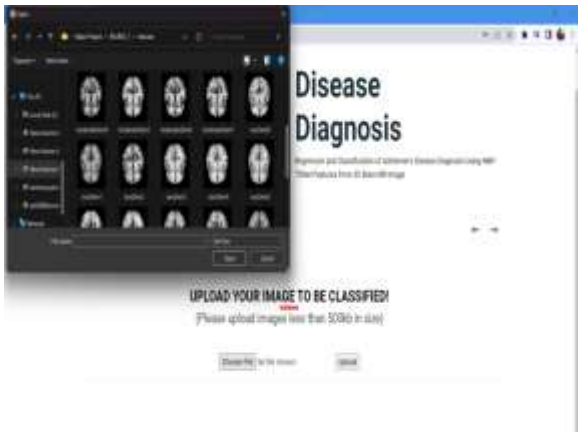


Fig.10: User input Sample



Fig.11: Predicting results Display View

VI. CONCLUSION

In this review, we present NMF-TDNet, a methodology in view of the organization construction of PCANet that conquers the limits of PCANet's enormous number of highlights and the PCA channels' information reliance.

NMF-TDNet employs layer-wise convolution to examine the information picture as opposed to PCA. The output layer neuron results are then utilized to construct a higher-demand tensor, and TD is utilized to make the last picture features by making the data more modest in dimensionality. In conclusion, we analyze Promotion and predict clinical scores by incorporating these characteristics into the SVM. The ADNI-1 and ADNI-2 datasets were subjected to class name isolation tests and clinical score estimations (MMSE, ADAS-11, and ADAS-13). In addition, separate classification names and foresee clinical scores (MMSE) were derived from the ADNI-1 and Desert garden datasets. In spite of the way that NMF-TDNet creates substantially less highlights than PCANet, the trial results uncover that using NMF-TDNet highlights as info brought about preferable execution over utilizing PCANet highlights as information.

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