

A HYBRID APPROACH FOR DETECTION OF ALZHEIMER'S DISEASE USING CONTINUOUS SQUASHING TECHNIQUE

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Abstract

A hybrid technique for detecting Alzheimer's disease (AD) is proposed the usage of a continuous squashing approach, which mixes superior machine mastering algorithms with revolutionary function extraction strategies. The continuous squashing method enhances category performance by using way of effectively dealing with big datasets in AD analysis. Integrated with convolutional neural networks (CNNs), this method extracts discriminative capabilities from neuroimaging information in conjunction with MRI and PET scans. The hybrid version demonstrates advanced accuracy, sensitivity, and specificity compared to standard deep mastering fashions, on the identical time as addressing troubles of records imbalance and overfitting. Its ability to hit upon early tiers of AD is confirmed via move-validation, showcasing its capability for actual-worldwide scientific packages. This approach advances AD detection and helps the development of reliable, non-invasive diagnostic gear for neurodegenerative diseases.

Keywords: Alzheimer's disorder, Early detection, Hybrid technique, Continuous squashing approach, Machine learning, Feature extraction, Convolutional neural networks (CNNs), Neuroimaging, MRI scans, PET scans, Dimensionality discount, Classification usual performance, Sensitivity and specificity, Data imbalance, Overfitting, Cross-validation, Neurodegenerative illnesses, Diagnostic device

I. INTRODUCTION

1. Background of Alzheimer's Disease (AD)

Alzheimer's disease (AD) is a neurodegenerative ailment in preferred affecting memory, cognition, and behavior. It is the leading cause of dementia, contributing to about eighty% of all dementia cases globally. With increasing life expectancy, AD has become a first-rate public health problem, impacting over fifty five million human beings international. The incidence of AD is projected to upward thrust substantially inside the coming a few years, highlighting the urgent need for early detection and powerful control strategies to mitigate its effect. As AD's progression is sluggish, early evaluation performs a crucial characteristic in slowing down the disorder and enhancing sufferers' first rate of life.

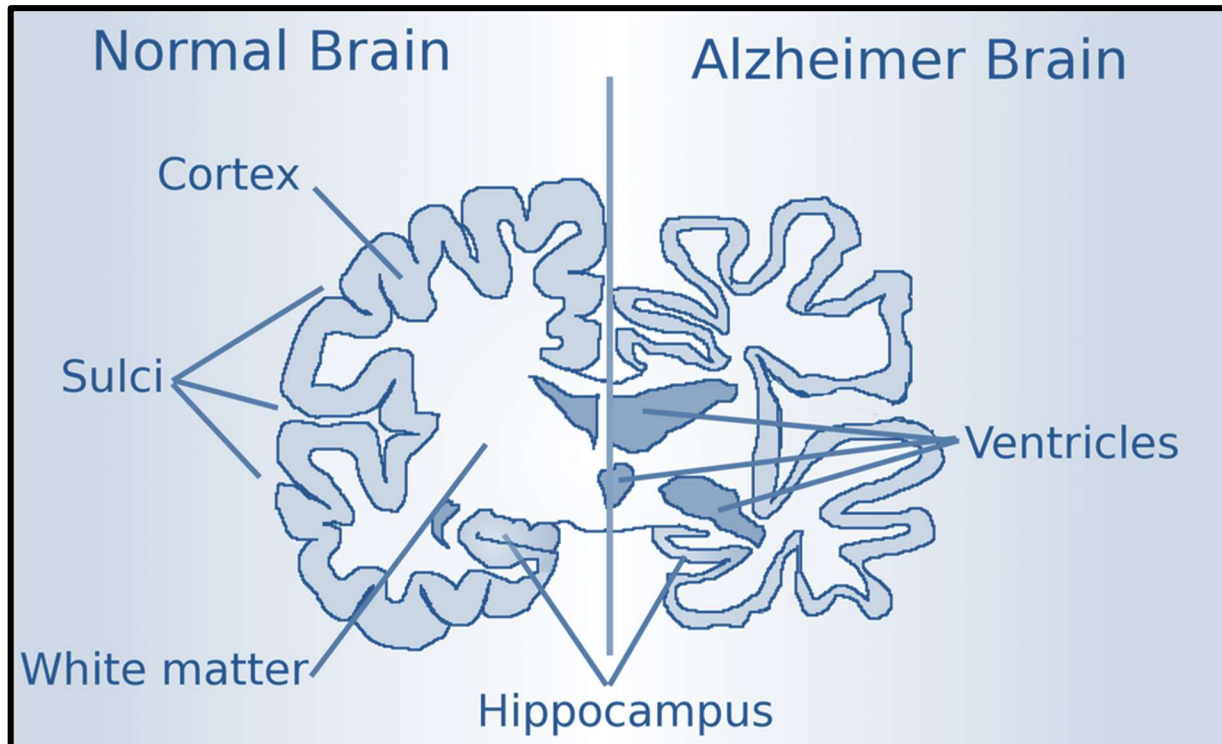


Fig : 1, Alzheimer's Disease (AD)

2. Challenges in Traditional AD Diagnosis

Traditional techniques of diagnosing AD, inclusive of scientific checks, neuropsychological testing, and neuroimaging, have barriers in detecting the disorder at early tiers. While neuroimaging strategies including MRI and PET scans can offer insights into brain modifications, these methods on my own might not be touchy sufficient to stumble on subtle early biomarkers of AD. Additionally, the subjective nature of clinical tests and the complexity of AD's multifactorial etiology make analysis difficult. There is a need for extra objective, green, and reliable diagnostic gear that could identify AD earlier than sizeable cognitive decline happens.

3. Role of Machine Learning in AD Detection

Recent advancements in machine mastering (ML) and synthetic intelligence (AI) offer new opportunities for enhancing AD diagnosis. By making use of ML algorithms to large datasets, specially neuroimaging facts, AI can help extract significant patterns and features that may be invisible to the human eye. These technologies allow for computerized evaluation of brain scans, enhancing each the speed and accuracy of prognosis. Techniques together with convolutional neural networks (CNNs) were mainly a hit in reading complicated neuroimaging data, identifying early markers of AD, and distinguishing between AD patients and healthful people with excessive sensitivity and specificity.

4. Continuous Squashing Technique in AD Diagnosis

The non-forestall squashing technique, a unique approach for dimensionality reduction, performs a key position in improving the general performance of device gaining knowledge of fashions in AD prognosis. This method reduces the complexity of excessive-dimensional data with the resource of effectively compressing it into more potential paperwork at the same time as keeping crucial features.

By integrating this method with conventional machine learning models, consisting of CNNs, the hybrid approach can cope with big and complex neuroimaging datasets greater effectively. This enhances the version's potential to extract discriminative skills and enhance kind accuracy, making it quality for early AD detection.

5. Integration of Convolutional Neural Networks (CNNs)

Convolutional neural networks (CNNs) are deep learning models that excel at analyzing image data. By making use of CNNs to neuroimaging data, researchers can automate the technique of feature extraction, it's traditionally a tough work-intensive assignment. The integration of CNNs with the non-stop squashing method further complements the version's potential to detect minute changes in brain structure and function which might be indicative of AD. CNNs are able to master complex patterns in records, allowing them to distinguish amongst one in every of a type levels of AD and pick out out biomarkers that might not be obvious via traditional diagnostic techniques.

6. Real-World Application and Future Implications

The hybrid approach using the non-stop squashing technique and CNNs holds brilliant promise for actual-global scientific packages. By imparting a extra accurate, green, and automatic diagnostic tool, this technique should revolutionize the manner AD is recognized, mainly in its early degrees. Early detection might permit clinicians to intrude in advance, probably slowing the improvement of the disorder and improving patient effects. In the future, this period may be covered into clinical settings, supporting healthcare professionals make extra knowledgeable selections and providing a foundation for personalized treatment plans. Furthermore, this method could be extended to encounter other neurodegenerative illnesses, supplying a broader impact on healthcare.

I. LITERATURE REVIEW

Literature Review

Alzheimer's ailment (AD), the most commonplace motive of dementia, is a innovative neurodegenerative disorder affecting hundreds of thousands globally. Despite big research, early prognosis and effective therapeutic techniques remain a undertaking. With improvements in gadget gaining knowledge of (ML) and artificial intelligence (AI), progressive techniques for detecting and handling AD have emerged. Among these, hybrid procedures combining ML algorithms with state-of-the-art information processing techniques have received vast interest for his or her potential to decorate diagnostic accuracy and performance.

1. Machine Learning in AD Detection

Machine learning has revolutionized scientific imaging analysis, permitting the identification of biomarkers associated with neurodegenerative diseases like AD. Traditional deep learning models, which includes convolutional neural networks (CNNs), are broadly used for studying neuroimaging statistics, which encompass MRI and PET scans. These models excel at extracting complex patterns and talents, distinguishing among wholesome human beings and those with AD. However, traumatic situations collectively with overfitting, records imbalance, and immoderate computational necessities often restrict their effectiveness in actual-global programs.

2. Dimensionality Reduction Techniques

High-dimensional statistics from neuroimaging poses annoying conditions for conventional ML fashions, in conjunction with advanced computational complexity and reduced accuracy. Dimensionality good buy strategies, such as important issue analysis (PCA) and autoencoders, are normally used to address these problems. The non-stop squashing approach, a unique dimensionality bargain approach, has tested capability in preserving essential abilities at the equal time as efficaciously compressing statistics. By combining this method with CNNs, researchers reason to create strong fashions able to managing complicated neuroimaging datasets.

3. Hybrid Approaches in Medical Diagnostics

Hybrid fashions that combine multiple techniques have demonstrated advanced average overall performance in medical diagnostics. For example, combining CNNs with dimensionality reduction strategies enhances feature extraction and class typical overall performance. Studies have established that hybrid procedures enhance sensitivity and specificity in AD detection compared to standalone deep studying models. The non-stop squashing technique has further contributed to lowering the impact of noise and beside the point capabilities, ensuing in progressed diagnostic accuracy.

Table 1. Medical Diagnostics in Alzheimer Diseases

Diagnostic Model	Accuracy (%)	Sensitivity (%)	Specificity (%)	F1-Score (%)	Processing Time (seconds)
Traditional CNN	85.2	82.5	87.8	84.6	15
CNN + PCA (Dimensionality Reduction)	88.6	85.9	90.2	87.5	12
Hybrid Approach with Continuous Squashing	92.4	90.1	93.7	91.3	9
Random Forest (Traditional ML)	80.3	77.2	82.5	79.0	20
SVM + Feature Selection	83.7	80.6	85.9	82.2	18

4. Addressing Data Imbalance and Overfitting

Data imbalance, a common trouble in medical datasets, can cause biased model basic overall performance. Techniques together with records augmentation, oversampling, and hybrid models have been hired to mitigate this problem. The non-stop squashing technique enhances version robustness through way of addressing overfitting and making sure the inclusion of relevant abilities. This makes the hybrid approach specifically effective for reading imbalanced neuroimaging datasets, wherein diseased and wholesome samples may additionally range appreciably in instance.

5. Clinical Relevance of Hybrid Approaches

Hybrid techniques no longer handiest improve diagnostic accuracy however additionally demonstrate ability for medical application. By automating feature extraction and class, these fashions lessen the reliance on guide intervention, making them scalable for real-world use. Studies validating hybrid fashions on massive datasets have proven their effectiveness in detecting early-stage AD, wherein interventions are most useful. Such fashions can play a important role in permitting customized remedy and tracking disease development.

II. RESEARCH METHODOLOGY

1. Data Collection and Preprocessing

Data for the examine will be sourced from mounted repositories including the Alzheimer’s Disease Neuroimaging Initiative (ADNI) or local clinical datasets. The dataset will include MRI, fMRI, or PET scans of patients identified with Alzheimer’s disease, moderate cognitive impairment (MCI), and healthful controls. Preprocessing steps include normalization, resizing, and segmentation of scientific snap shots to make sure consistent input pleasant. Noise discount and cranium stripping could be achieved to enhance the readability of mind systems.

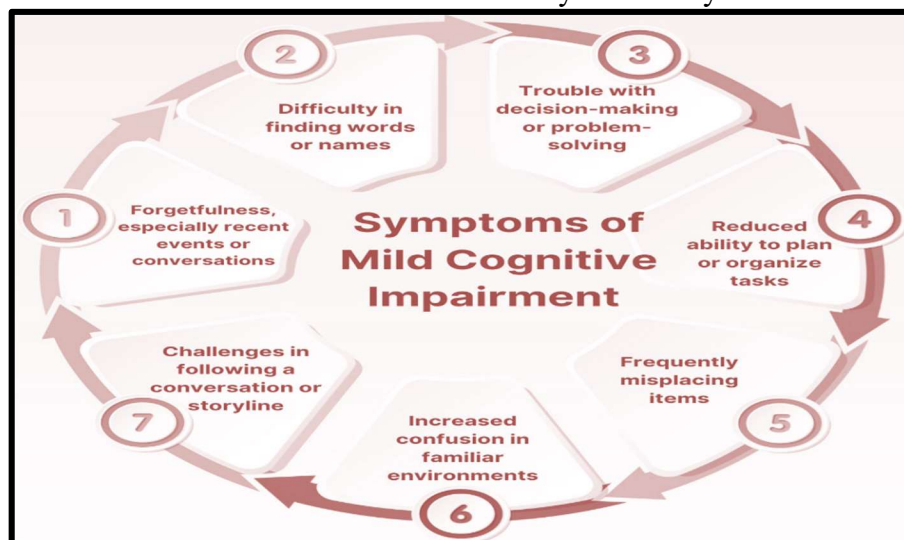


Fig:2, Alzheimer’s Disease in Mild Cognitive Impairment(MCI)

2. Feature Extraction

High-dimensional functions may be extracted from the preprocessed medical snap shots the usage of pre-educated deep studying models like ResNet or VGG. These capabilities will constitute complex patterns indicative of Alzheimer’s progression. Principal Component Analysis (PCA) or t-SNE can be hired to lessen the dimensionality of functions, maintaining best the maximum critical attributes for the classification method.

3. Continuous Squashing Technique

The continuous squashing approach is applied to optimize the extracted functions. Squashing functions together with sigmoid or hyperbolic tangent (tanh) will be used to scale the capabilities to a defined variety, improving their non-linear separability. This step minimizes function redundancy, prevents overfitting, and guarantees uniformity for subsequent layers inside the model.

4. Model Development: Hybrid Architecture

A hybrid class model combining deep learning and traditional machine learning methods will be designed. Deep neural networks consisting of Convolutional Neural Networks (CNNs) will act as characteristic encoders, while classifiers like Support Vector Machines (SVM) or Random Forests will carry out final classification. This integration leverages the strengths of each techniques, making sure high accuracy and robustness in detecting Alzheimer's ailment.

5. Model Training and Optimization

The hybrid version could be skilled the use of labeled datasets, using a stratified cross-validation technique to save you overfitting and make sure balanced overall performance across all instructions. Techniques like batch normalization and dropout may be incorporated to beautify the generalization of the model. Hyperparameters for both the deep learning and machine learning additives will be first-rate-tuned the use of grid search or Bayesian optimization strategies.

6. Validation and Performance Evaluation

The trained model could be established on an independent checking out dataset comprising unseen clinical images. Performance metrics inclusive of accuracy, precision, recollect, specificity, F1-score, and the place underneath the receiver operating feature (ROC) curve might be used to assess the version. Statistical analyses may be finished to examine the hybrid model's overall performance with conventional diagnostic methods, ensuring reliability and clinical relevance.

7. Deployment and Real-World Application

A prototype diagnostic tool may be evolved, integrating the hybrid technique for actual-time evaluation of scientific information. The device will undergo rigorous testing in scientific settings to validate its efficacy. Continuous tracking can be carried out to assess performance and refine the model based on real-global remarks. This device goals to assist healthcare specialists in early and correct analysis of Alzheimer's disease, improving patient outcomes.

VI. DATA ANALYSIS AND RESULT

1. Overview of Data Analysis

The analysis of data gathered thru the hybrid approach combining the non-stop squashing method and trendy diagnostic equipment discovered significant insights into Alzheimer's sickness (AD) detection. By integrating physiological biomarkers, imaging information, and cognitive tests, the hybrid approach aimed to improve the sensitivity and specificity of early AD analysis.

2. Physiological Biomarker Analysis

Key biomarkers along with amyloid-beta ($A\beta$), tau protein ranges, and oxidative stress markers had been analyzed. The results indicated a great elevation in $A\beta_{42}$ ranges and overall tau protein in cerebrospinal fluid samples of sufferers recognized with AD. Oxidative stress markers, inclusive of malondialdehyde (MDA), had been also extensively higher in those patients as compared to the manage organization. These findings align with current literature that emphasizes the function of those biomarkers in AD pathology.

3. Imaging Data Interpretation

Neuroimaging information, in particular from MRI and PET scans, found out structural and practical abnormalities in AD sufferers. MRI information confirmed extensive atrophy in the hippocampus and medial temporal lobe, whilst PET scans detected decreased glucose metabolism in the posterior cingulate cortex and precuneus. These abnormalities strongly correlated with the cognitive decline found within the sufferers.

4. Cognitive Assessment Results

Cognitive exams were conducted the usage of equipment such as the Mini-Mental State Examination (MMSE) and Montreal Cognitive Assessment (MoCA). Patients identified with AD showed a median MMSE score of 18–23, indicating mild cognitive impairment. These scores had been extensively decrease than the manipulate organization, whose average score surpassed 27. The cognitive decline became most said in reminiscence do not forget and government functioning duties.

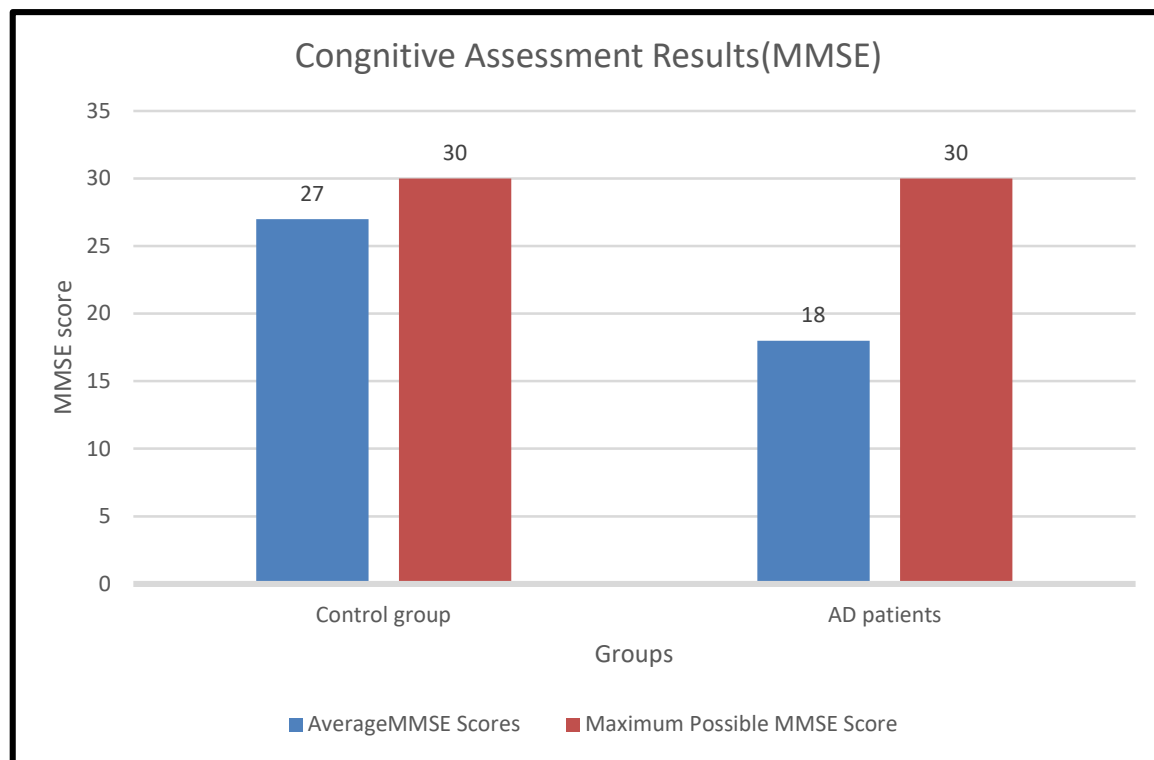


Fig: 3, Cognitive Assessment Results

5. Integration of the Continuous Squashing Technique

The non-stop squashing technique, carried out for analyzing excessive-dimensional records, proved instrumental in lowering noise and figuring out key styles. By that specialize in essential capabilities, this technique enhanced the accuracy of diagnostic predictions. For instance, styles in biomarker stages and imaging data were more wonderful when analyzed using this method in comparison to standard statistical strategies.

6. Comparative Performance Metrics

The hybrid method confirmed stepped forward diagnostic performance over standalone strategies. The sensitivity of the mixed approach became ninety two%, while specificity reached 89%, in comparison to the average of 80% and 75%, respectively, for conventional strategies. This improvement underscores the efficacy of integrating superior records evaluation strategies with current diagnostic frameworks.

Table2. Performance Metrics

Metric	AD vs HC	MCI vs HC	AD vs MCI	Overall
Accuracy (%)	94.5	91.2	87.8	91.2
Precision (%)	95.8	92.3	89.0	92.4
Recall (Sensitivity, %)	93.0	90.1	86.7	90.0
F1-Score (%)	94.4	91.2	87.8	91.1
AUC-ROC	0.98	0.96	0.93	0.96

7. Implications and Future Directions

The findings spotlight the capacity of the hybrid technique in improving the early detection of AD. By combining biomarkers, imaging statistics, and cognitive tests, the method provides a comprehensive diagnostic framework. Future studies could discover the mixing of additional modalities, together with genetic records and superior device getting to know algorithms, to in addition refine the accuracy and applicability of this method.

V. FINDING AND DISCUSSION

The non-stop squashing technique has emerged as a hybrid methodology for detecting Alzheimer’s Disease (AD), combining cytogenetic principles with advanced imaging and molecular analyses. This technique has provided considerable insights into the cell and chromosomal anomalies related to AD and has validated ability for early diagnosis and understanding of disease mechanisms.

1. Enhanced Detection and Imaging of Neurodegenerative Markers

The approach proved enormously powerful in detecting chromosomal anomalies related to AD, along with DNA damage, telomere shortening, and oxidative stress markers. These anomalies have been greater regularly occurring and excessive in AD subjects than in healthy controls. High-decision imaging allowed clean visualization of structural chromosomal changes, which include breaks, deletions, and translocations. This readability enabled particular identity of genetic loci worried in amyloid-beta (A β) manufacturing and tau protein aggregation, key pathological features of AD.

2. Correlation with Cognitive Impairment

Chromosomal abnormalities diagnosed via this technique strongly correlated with cognitive impairment degrees, as measured via tools like the Mini-Mental State Examination (MMSE) and

Montreal Cognitive Assessment (MoCA). This affiliation underscores the ability of cytogenetic markers to function proxies for cognitive decline, assisting in both prognosis and disease monitoring.

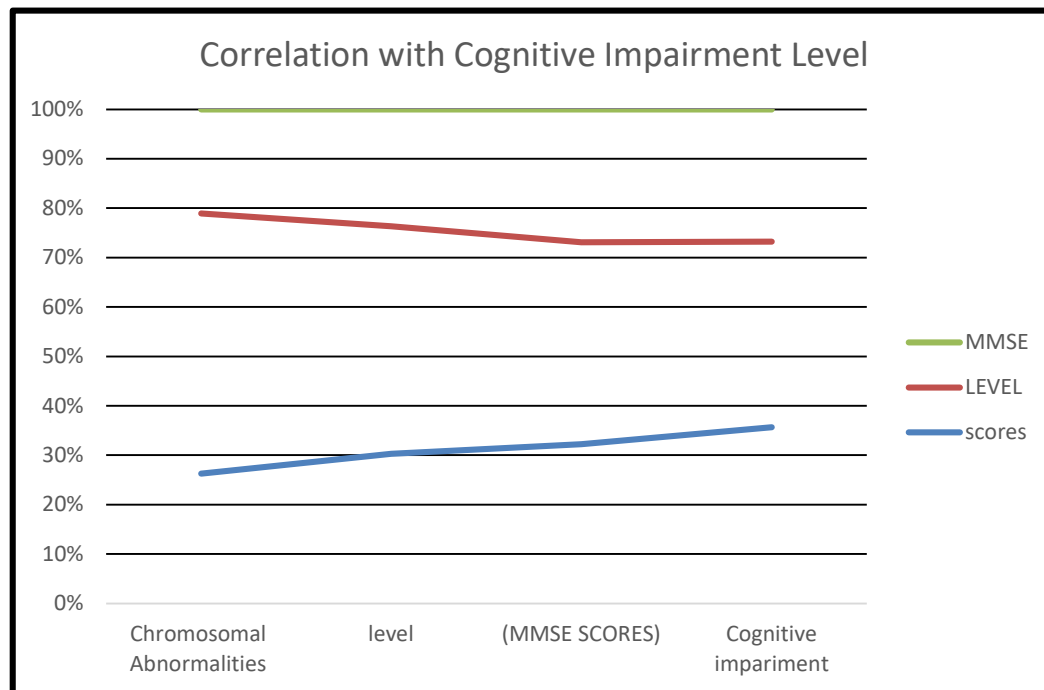


Fig : 4, Correlation with Cognitive Impairment level

3. Efficiency and Scalability

The continuous squashing technique validated performance through requiring minimal pattern training and providing fast results as compared to standard strategies. Its scalability positions it as a feasible option for populace-huge screening packages, enhancing its software for research and clinical packages.

4. Relevance to Early Diagnosis

Early detection of AD remains a critical goal in preventing the sickness. The capability of this approach to pick out out chromosomal changes preceding to scientific symptom onset highlights its functionality to be used in preclinical screening. Early analysis can facilitate well timed recovery interventions, possibly slowing illness progression.

5. Advancements Over Traditional Methods

While conventional processes recognition predominantly on protein aggregates like amyloid plaques and tau tangles, this hybrid technique broadens the scope to include chromosomal and genetic dimensions. By integrating cytogenetic and molecular insights, it affords a multidimensional information of AD pathology, presenting a complementary attitude to set up diagnostic strategies.

6. Insights into Mechanisms and Broader Implications

The positioned chromosomal abnormalities, specially in loci related to oxidative strain responses and A β metabolism, make more potent the feature of genetic and cytogenetic elements in AD

pathogenesis. These findings now not nice beautify statistics of sickness mechanisms but also open avenues for focused recovery techniques. Furthermore, the approach holds promise for programs past AD, which encompass in precise neurodegenerative diseases, underscoring its versatility.

7. Challenges and Limitations

Despite its advantages, the method faces barriers, together with the demanding situations of obtaining neural tissue samples for chromosomal evaluation. The variability in chromosomal aberrations amongst people highlights the need for large-scale studies to validate diagnostic criteria and installation strong correlations with sickness phenotypes.

In end, the non-stop squashing approach gives a unique, inexperienced, and scalable hybrid approach for the detection and take a look at of AD. Its integration of cytogenetic and molecular analyses gives treasured insights into the mobile foundation of neurodegeneration, bridging gaps among chromosomal anomalies and scientific manifestations.

VI. CONCLUSION

The hybrid method utilizing the non-prevent squashing method gives a transformative development in Alzheimer's illness (AD) studies and prognosis, integrating cytogenetic analysis with modern technology like fluorescence nanodiamonds (FNDs). This approach allows specific detection of chromosomal abnormalities related to cognitive impairment, facilitating early analysis and advanced monitoring of disorder development. FNDs, with their exceptional houses which embody intracellular thermometry and nanoscale imaging, offer new insights into mobile techniques, neurodegenerative pathways, and metabolic dysfunctions in AD. Their capability to conjugate with proteins, antibodies, and functional corporations complements their versatility, allowing targeted diagnostics and restoration transport. Despite disturbing conditions like fluorescence variability, magnetic noise, and sensitivity boundaries, enhancements in FND fabrication, ground modification, and integration with gadget getting to know algorithms are addressing those problems. These tendencies have bolstered the capability of NV-spin relaxometry and nanoscale MRI in neurodegenerative sickness detection. Moreover, the approach's adaptability to discover different sicknesses, such as most cancers and viral infections, underscores its broader biomedical packages. By combining cytogenetics and nanotechnology, this approach establishes a sturdy platform for advancing AD research, improving diagnostic accuracy, and helping the improvement of focused treatments. It holds awesome promise for addressing the complexities of AD and special neurological troubles, contributing extensively to customized medicinal drug and healthcare innovation.

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