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Research on Mental Health Prediction using ML

Asst. Prof. Ms. Jaishree Wankhede[1], Shriram Phatale[2], Pranjay Dharkar[3], Rohan Ramteke[4], Ishika Rawlani[5], Dimpal Mathre[6]
Dept. Artificial Intelligence & Data Science, P.C.E, Nagpur

Abstract: ***This review paper explores the application of Artificial Intelligence (AI) and Machine Learning (ML) techniques in mental health detection and therapy. From early diagnosis to personalized interventions, AI has introduced scalable, accessible solutions for mental health challenges. By investigating existing strategies, such as facial expression investigation and chatbots, and tending to challenge like moral concerns and information quality, this paper emphasizes AI's transformative part in mental wellbeing care.***
Keywords: ***Mental Health, Artificial Intelligence, Machine Learning, Diagnosis, Therapy***

I. INTRODUCTION

Mental sickness is progressively recognized as a worldwide scourge, influencing individuals over all socioeconomics. Agreeing to the World Wellbeing Organization (WHO), one in four people will encounter a mental or neurological clutter at a few points in their lives. Depressive clutters have risen as a critical concern, anticipated to gotten to be the moment driving supporter to the worldwide infection burden by the early 2020s, outperformed as it were by ischemic heart illness. Despite the raising predominance, the accessibility of qualified mental wellbeing experts has not kept pace, making a considerable hole within the openness of mental health care

The method of diagnosing mental wellbeing clutters is intrinsically complex. It ordinarily includes a combination of organized interviews, an appraisal of restorative history, physical examinations, and mental assessments. These strategies point to recognize mental wellbeing issues from indications stemming from physical conditions. In any case, covering indications over different mental wellbeing disarranges can lead to symptomatic challenges, counting potential misdiagnosis. This issue gets to be indeed more articulated when diagnosing mental well-being conditions in children, requiring cautious, exact approaches.

In recent years, Artificial Intelligence (AI) has emerged as a transformative tool in mental health diagnosis and care. AI systems, powered by Machine Learning (ML), have demonstrated the capability to analyze data, mimic human reasoning, and make predictions even in the presence of incomplete or ambiguous information. Such systems are proving invaluable in assisting professionals with diagnostic processes.

This paper will identify and assess different ML techniques for diagnosing mental health conditions, considering Attention Problems, Academic Difficulties, Anxiety Disorders, Attention Deficit Hyperactivity Disorder (ADHD), and Pervasive Developmental Disorders (PDD). The symptoms and behavioral pattern data, along with other forms of assessment results, create a strong structure through AI models toward truly accurate and efficient diagnostics in mental health—an enabler for innovative and scalable solutions in mental health care.

II. LITERATURE REVIEW

Research in machine learning in mental health diagnosis started in the 1980s. DTREE is an expert system that diagnoses DSM-IV Axis I disorders using Decision Tree techniques. INTERNIST/AUTOSCID is a computerized Structured Clinical interview for DSM-IV Axis II personality disorders. An expert system called MILP (Monash Interview for Liaison Psychiatry) developed by Yap R.H. and Clarke D.M. uses constraint-based reasoning to carry out systematic diagnoses of mental disorders based on DSM-III-R, DSM-IV, and ICD-10.

In [7], Masri R.Y., and Jani H.M. proposed an Expert System for Mental Health Diagnosis, which helps psychologists diagnose and deal with their mentally ill patients. They employed three different artificial intelligence approaches for diagnosis and treatment recommendation: Rule-Based Reasoning, Fuzzy Logic, and Fuzzy-Genetic Algorithm. Luxton, David D.D. [8] examined the uses of artificial intelligence within psychological practice. Current and future trends and their impact are discussed here.

Razzouk D., et al [9] created a choice bolster framework for the conclusion of schizophrenia clutters with an exactness of 66-82%. Chattopadhyay S., et al. [10] created a neurofuzzy approach to review grown-up misery.

The directed (Back Proliferation Neural Arrange BPNN and Versatile Network-based Fluffy Induction Framework ANFIS) and unsupervised (Self-Organizing Outline) neural net learning approaches have been utilized and after that compared. It was watched that ANFIS being a cross breed framework performed much way better than BPNN classifier.

Basavappa S.R. et al. [12] utilized profundity to begin with look strategy with in reverse look procedure to analyze sadness or dementia. They created an master framework utilizing the patient's behavioral, cognitive, passionate indications and comes about of neuropsychological appraisals. Rahman, Rashedur M. and Farhana Afroz [13] compared different classification procedures (Bayesian Arrange, Multilayer Perceptron, Choice Trees, Single Conjunctive Run the show Learning and Fluffy Induction Frameworks and Neuro-Fuzzy Deduction Framework) utilizing diverse information mining devices like WEKA, TANAGRA and MATLAB for diagnosing diabetes.

David Gil A. and Magnus jhonson B. in [14], proposed a framework based on Fake Neural Networks(ANN) and Back Vector Machines(SVM) that analyze Parkinson's Infection. The framework has appeared an increment in precision and a diminish in fetched. In [15], Gomuła, Jerzy et al., attempted to discover productive strategies for classification of MMPI profiles of patients with mental clutter. They recognized that Quality Expansion Approach makes strides classification precision within the case of discretized information.

Flavio Luiz Seixas and Bianca Zadrozny in [16] proposed a Bayesian Organize (BN) Choice Show for determination of dementia, Alzheimer's malady and Gentle Cognitive Disability. The BN model was considered because it is well suited for representing uncertainty and causality. Organize parameters were evaluated employing a directed learning calculation from a dataset of genuine clinical cases.

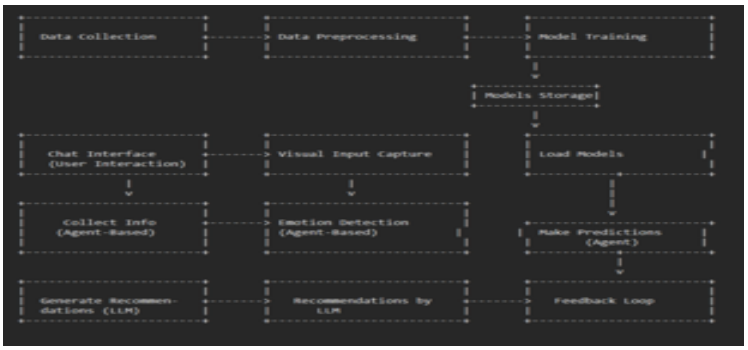
Anchana Khemphila and Veera Boonjing in [17] utilized Multi-Layer Perceptron (MLP) with Back Engendering Learning to analyze Parkinson's illness successfully with diminished number of qualities. Data Pick up of all properties is utilized as a degree to diminish the number of properties. Pirooznia, Mehdi, et al., in [18] utilized information mining approaches for Genome-wide Affiliation of Temperament Disarranges. It was distinguished that straightforward polygenic score classifier performed superior than others and it was moreover found that all the classifiers performed more regrettable with little number of Single Nucleotide Polymorphisms within the brain communicated set compared to entire genome set.

In [5], Kipli, Kuryati, Abbas Z. Kouzani, and Matthew Joordens recognized misery from basic MRI looks to analyze the mental wellbeing of patients. They explored exhibitions of four Include Determination calculations, specifically, OneR, SVM, Data Pick up (IG) and ReliefF. At last, they concluded that the SVM Evaluator in combination with Desire Maximization (EM) classifier and the IG evaluator in combination with Arbitrary Tree Classifier have accomplished the most elevated precision. It had moreover been found that the little basic sizes restrain the capacity to draw firm conclusions.

Dabek, Filip, and Jesus J. Caban [19] created a Neural Organize (NN) Demonstrate with an precision of 82.35% for anticipating the probability of creating mental conditions such as uneasiness, behavioral clutters, misery and post-traumatic stretch clutters. In [20], Tawseef Ayoub Shaikh, compared the execution of Manufactured Neural Systems, Choice Tree and Naïve Bayes in anticipating Parkinson's infection and Essential Tumor Illness and found that the exactness is tall in ANN for foreseeing Parkinson's infection and Naïve Bayes for Essential Tumor Malady. Exactness of Choice Tree and Naïve Bayes have advance progressed after lessening the estimate of include set by applying Hereditary Calculation to the real information set.

III. METHODOLOGY

The methodology for diagnosing mental health issues using AI and Machine Learning involves a systematic process that integrates data collection, pre-processing, model training, and deployment. This section details the steps followed in building a robust AI-powered system for mental health detection.



A. Data Collection

The foundation of this system is a comprehensive dataset comprising clinical and behavioural data. The dataset includes:

1) Demographic Information: Age, gender, family history of mental health issues.
2) Symptoms and Behavioural Indicators: Mood swings, lack of concentration, fatigue, and compulsive behaviours.
3) Assessment Scores: Standardized tests such as CBCL (Child Behaviour Checklist) or GAD (Generalized Anxiety Disorder) scores.
4) Clinical Diagnoses: Labelled data indicating the presence or absence of specific mental health conditions.

B. Data Pre-processing

Raw data is pre-processed to ensure consistency and usability for ML models. The key steps include:

1) Feature Selection: Identifying relevant attributes using techniques like Best First Search to eliminate redundant data.
2) Normalization: Scaling data to a standard range for uniformity across attributes.
3) Handling Missing Data: Filling gaps using statistical imputation techniques or removing incomplete records.
4) Class Balancing: Addressing imbalances in datasets to ensure unbiased model training.

C. Model Selection and Training

Various ML algorithms were evaluated for their suitability in detecting mental health conditions. The models selected include:

1) Multilayer Perceptron (MLP): A neural network model for non-linear relationships.
2) Random Forest: An ensemble learning method for high-accuracy classification.
3) Logistic Regression: A statistical model for binary classification.
4) Support Vector Machine (SVM): Effective for high-dimensional data analysis. The data was split into training and testing subsets, typically in an 80:20 ratio. The training subset was used to build models, and hyper parameter tuning was performed to optimize performance.

D. Demonstrate Assessment

Each model's execution was assessed utilizing measurements such as:

1) Accuracy: Percentage of correctly classified instances.
2) Precision and Recall: Metrics for evaluating true positive and false positive rates.
3) ROC-AUC Curve: Analysing the trade-off between sensitivity and specificity.

E. Deployment

The best-performing model, based on evaluation metrics, was integrated into a web-based system for real-time detection.

1) Web Application: A user-friendly interface for data input and visualization of results, developed using frameworks like React.
2) Flask API: Acts as a bridge between the web application and the ML model for real-time predictions.
3) System Workflow: Users input their details through a web interface. The data is processed and analysed by the ML model, and results are displayed to the user.

F. Continuous Improvement

The system is designed for iterative improvement. New data can be incorporated into the model, allowing retraining and enhanced accuracy over time. Feedback from mental health professionals and end-users is leveraged to refine the diagnostic approach.

IV. CONCLUSION

AI-powered frameworks hold monstrous guarantees for changing mental wellbeing care by advertising convenient conclusion, versatile treatment apparatuses, and personalized mediations. Facial acknowledgment innovations, chatbots, and prescient models have as of now appeared empowering comes about in identifying and tending to mental wellbeing challenges. In any case, for broad appropriation, it is crucial to address moral concerns, guarantee vigorous information quality, and create models that are comprehensive and versatile to different populaces.

Long-term mental health care lies at the crossing point of innovation and sympathy, where AI frameworks help instead of supplanting human experts, guaranteeing that mental wellbeing bolster is both available and viable.

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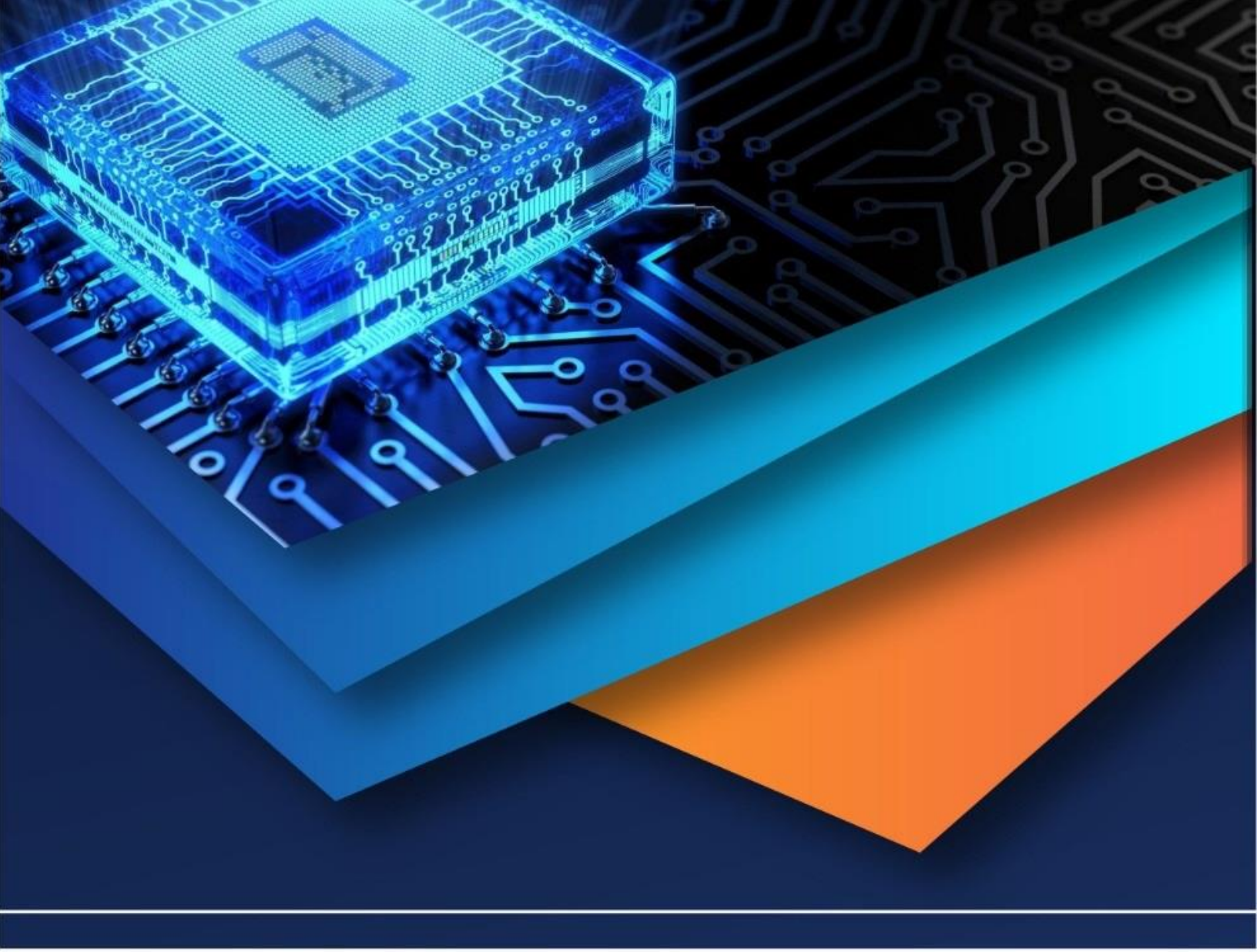

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