



Epileptic Seizure Recognition using Machine Learning

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Abstract: Epilepsy, a severe neurological disorder, is identified by analyzing intricate brain signals generated by interconnected neurons, often monitored through EEG and ECoG. These signals, characterized by complexity, noise, and non-linearity, pose significant challenges for seizure detection. However, recent strides in machine learning have facilitated the development of robust classifiers capable of effectively analyzing EEG and ECoG data. By leveraging these advancements, researchers can accurately detect seizures and extract pertinent patterns, thereby aiding in the diagnosis and management of epilepsy. Machine learning techniques empower clinicians to uncover valuable insights into the condition, ultimately enhancing patient care and treatment strategies. The integration of machine learning with EEG and ECoG analysis holds promise for advancing our understanding of epilepsy and improving patient outcomes.

Keywords: Seizure detection, data preprocessing, training the model, EEG signals, LSTM model, machine learning.

I. INTRODUCTION

Epilepsy is a condition of the brain that is characterized by recurrent seizures, which can occur without any apparent cause. It is the fourth most common neurological disorder globally, affecting people of all ages, races, and ethnic backgrounds. Seizures are caused by surges of abnormal electrical activity in the brain, leading to various manifestations such as confusion, muscle jerks, or unconsciousness. The diagnosis of epilepsy typically occurs after a person experiences two or more unprovoked seizures. Treatment often involves anti-seizure medication to manage and reduce the occurrence of seizures. Understanding epilepsy is crucial as it can impact safety, relationships, work, and various aspects of daily life. The field of epilepsy has seen significant advancements in treatment and research, aiming to improve outcomes and reduce the social stigma associated with the condition. Early diagnosis, appropriate treatment, and ongoing support are essential for individuals living with epilepsy to lead full and healthy lives. Additionally, understanding the distinction between provoked seizures, non epileptic events due to other medical problems, and psychogenic non epileptic seizures is essential in providing proper care and management for individuals experiencing seizures. Epileptic seizure recognition is a critical aspect of managing epilepsy, a neurological disorder characterized by abnormal brain activity leading to seizures. The detection of epileptic seizures plays a vital role in improving the quality of life for individuals affected by epilepsy. Various techniques, such as electroencephalography (EEG), have been employed to evaluate brain activity during seizures, aiding in the automatic detection of seizure.

II. LITERATURE SURVEY

[1] The paper introduces a learning framework using Gated Recurrent Unit (GRU) RNNs for seizure detection, addressing the critical need for automated systems due to the unpredictable nature of seizures. Focusing on a 3-class classification problem, the study collates EEG data into healthy, inter-ictal, and ictal states, employing 51 sub-segments to tackle the challenge of training on long-temporal sequences. This approach enables effective modeling of temporal sequences, enhancing the potential for accurate seizure detection and early warning systems.

[2] The paper provide a neonatal EEG dataset with seizure annotations for early diagnosis. EEG files are de-identified and annotated by three experts using Nicolet Reader software, with seizure onset and duration provided in MATLAB and CSV formats. Code availability for reading EEG files and montages aids accessibility for research.

[3] The paper introduces a review of epileptic seizure detection methods, focusing on EEG and ECoG data and employing statistical features and machine learning classifiers. They emphasize the superior performance of decision forest classifiers compared to others and discuss their effectiveness in seizure localization. This work aids in refining detection techniques and exploring brain lobe localization, providing valuable insights for researchers in the field.



[4] The paper present an automated epileptic seizure detection system utilizing multi-channel EEG analysis with features like PSD, entropy, and Teager energy, alongside a Multilayer Perceptron Neural Network (MLPNN) classifier. Preprocessing removes noise, while descriptive analysis confirms feature suitability, showing distinct band differences between normal and epileptic EEG signals. Simulation results exhibit high sensitivity (97.8%), specificity (96.4%), and low false detection rate (1 h^{-1}) with multi-features, indicating suitability for real-time seizure recognition. They develop a MATLAB-based graphical user interface called 'Aepitect' for automated biomarker extraction, aiming for faster and more accurate diagnosis, facilitated by the decision forest classifier, deemed most effective.

[5] The paper propose a method using Support Vector Machines (SVMs) to build patient-specific seizure detectors based on scalp EEG signals. They compare detector performance using feature vectors assembled from EEG alone versus EEG combined with ECG data. Results show improved performance with the addition of ECG information, reducing mean latency from 4.2 to 2.7 seconds, detecting all seizures, and decreasing false detections from 9 to 5 per 24 hours. This suggests the efficacy of incorporating ECG data for more accurate epileptic seizure detection.

III. SCOPE AND METHODOLOGY

Scope

The scope of a project focused on epileptic seizure recognition using machine learning encompasses various aspects crucial for the successful development and implementation of seizure detection algorithms. The scope includes Data Collection and Preprocessing, Feature Extraction and Selection, Model Development, Model Training and Evaluation, Real-Time Implementation, Clinical Validation, User Interface Development. Ultimately, the project aims to contribute to real-time seizure detection systems that can enhance patient monitoring and medical intervention.

Methodology

Machine Learning, a cornerstone of contemporary Information Technology, employs Artificial Intelligence (AI) to imbue systems with autonomous learning capabilities, allowing them to refine operations based on experiential knowledge rather than explicit programming. Within this domain, Neural Networks stand out for their versatility, finding applications in various security contexts, including face recognition, object detection, and activity classification. In addressing the challenge of video classification into crime or safety categories, a systematic methodology is implemented. Initially, 15 frames are extracted from the input video and standardized to a size of 200x200 pixels. These frames undergo preprocessing to eliminate backgrounds and convert them to grayscale, facilitating streamlined visual data analysis. Leveraging the Inception V3 pre-trained model, transfer values are extracted from each frame, capturing essential features pivotal for classification tasks.

These transfer values serve as inputs to an LSTM (Long Short-Term Memory) model, a specialized neural network architecture adept at analyzing temporal patterns and facilitating predictive analysis. By harnessing LSTM's capabilities, the system gains insight into nuanced temporal dynamics inherent in video data, enhancing its ability to discern between potentially criminal activities and benign actions. Through the integration of advanced machine learning techniques, including feature extraction and predictive modeling, the approach optimizes computational efficiency while ensuring robust classification accuracy. Ultimately, this fusion of methodologies holds promise for improving threat detection and enabling proactive security measures across diverse surveillance contexts.

IV. SYSTEM ARCHITECTURE

To implement epileptic seizure detection using LSTM, begin by preprocessing the EDF dataset, converting it into CSV format. Split the data into chunks of 178 data points for 1 second, resulting in 23 chunks per 4097 data points, with labels corresponding to different EEG states. Next, train the LSTM model using these preprocessed CSV files, where the input features are the 178-dimensional EEG recordings and the output is the category label indicating seizure presence or absence. Once trained, input a new EDF file for testing, preprocess it similarly,

and use the trained LSTM model to predict whether it contains a seizure or not. This approach leverages LSTM's ability to learn temporal dependencies in sequential data, allowing it to effectively detect patterns indicative of epileptic seizures from EEG recordings.

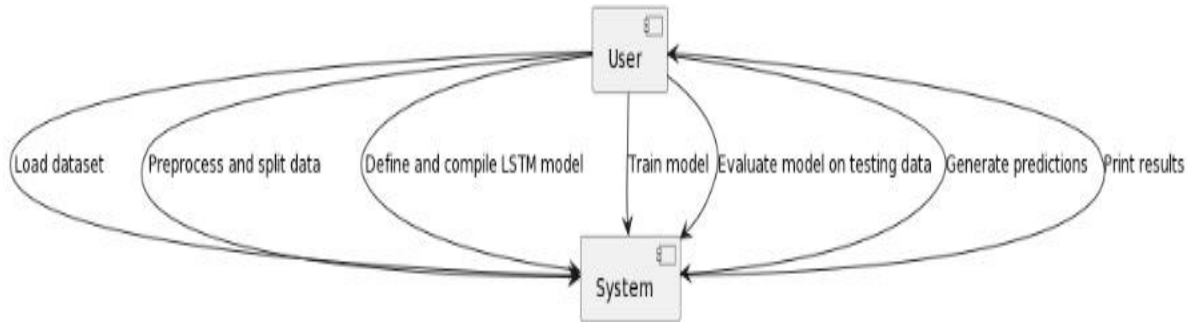


Fig. 1 System architecture

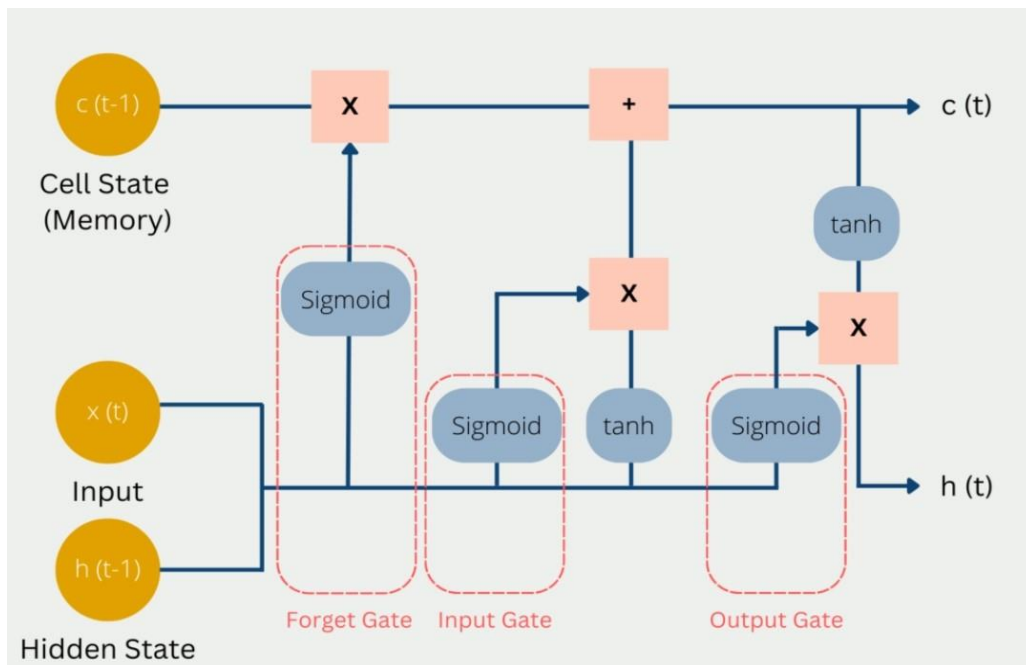


Fig.2 LSTM Architecture

V. CONCLUSION

The impact of this project utilizing machine learning, particularly LSTM, for epileptic seizure recognition includes enhanced accuracy, improved efficiency through automation, better patient outcomes, potential integration with wearable technology, and advancements in medical diagnostics. This project is of great practical importance in the modern world where it, has its potential to revolutionize epilepsy diagnosis and treatment. By leveraging advanced machine learning techniques like LSTM networks, this project offers enhanced accuracy in detecting seizures from EEG signals, automation for efficient diagnosis, improved patient outcomes through timely interventions, and the possibility of integration with wearable technology for real-time monitoring.

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