

## Automated Seizure Onset Detection and Prediction Using EEG: A Comprehensive Model with Multiscale Analysis and Machine Learning

Hassan Aridy <sup>1\*</sup>, Rayan Abu-Hanifa <sup>1</sup>, Jabbar Thaqib <sup>2</sup>, Naser Al-athiry <sup>3</sup>

<sup>1</sup> New Technologies Department, Al Iraqia University, Baghdad, Iraq

<sup>2</sup> Electrical and Electronic Engineering Department, Al Iraqia University, Baghdad, Iraq

<sup>3</sup> University of Information Technology and Communications, Baghdad, Iraq

(Received: 24<sup>th</sup> June 2021; Accepted: 2<sup>nd</sup> September 2021; Published on-line: 8<sup>th</sup> September 2021)

### Abstract

Accurate and timely detection of seizure onset is crucial for effectively managing epilepsy. This paper presents a novel and fully specified model for automated seizure onset detection and prediction based on electroencephalography (EEG) measurements. The model is evaluated using two widely recognized EEG databases, namely Freiburg (intracranial EEG) and CHB-MIT (scalp EEG), to assess its performance against state-of-the-art models. The proposed model incorporates four key components to enhance its effectiveness: (1) multiscale principal component analysis for EEG de-noising to improve signal quality, (2) EEG signal decomposition using empirical mode decomposition, discrete wavelet transform, or wavelet packet decomposition to capture relevant signal characteristics, (3) statistical measures for extracting informative features, and (4) machine learning algorithms for accurate classification. Experimental results demonstrate the superior performance of our model in comparison to existing approaches. It achieves an impressive overall accuracy of 100% in distinguishing between ictal and interictal EEG for both the Freiburg and CHB-MIT databases. Furthermore, the model exhibits high accuracy in seizure onset prediction, successfully discriminating between inter-ictal, pre-ictal, and ictal EEG states with an accuracy of 99.78% and distinguishing between inter-ictal and pre-ictal EEG states with an accuracy of 99.72%. Importantly, the proposed model's versatility extends beyond seizure detection and prediction. It can be applied to other classification tasks involving bio-signals, such as electromyography (EMG) and electrocardiography (ECG). Thus, our model holds promise as a comprehensive tool for various biomedical signal analysis applications.

**Keywords:** Electroencephalography, Seizure onset detection, Seizure onset prediction, EEG de-noising, Feature extraction, Machine learning.

### I. Introduction

Epilepsy is a neurological disorder characterized by recurrent seizures, and timely detection of seizure onset plays a vital role in its management. Electroencephalography (EEG) is a widely used technique for monitoring and analyzing brain activity, offering valuable insights into seizure patterns. This study aims to develop a comprehensive model for automated seizure onset detection and prediction using EEG data [1]. The proposed model combines several key components to enhance its performance and accuracy. Firstly, multiscale principal component analysis is employed for EEG de-noising, enabling the removal of noise and artifacts to improve signal quality. Secondly, the model incorporates various signal decomposition

techniques, such as empirical mode decomposition, discrete wavelet transform, or wavelet packet decomposition, to extract meaningful features from the EEG data. These decomposition methods enable identifying and extracting relevant signal characteristics at different scales [2]. Statistical measures are then applied to the decomposed signals to extract informative features that capture important patterns associated with seizure onset. Finally, machine learning algorithms are utilized for accurate classification and prediction based on the extracted features [3-4].

To evaluate the effectiveness of the proposed model, two benchmark EEG databases, Freiburg (intracranial EEG) and CHB-MIT (scalp EEG), are employed. The model's performance is compared against state-of-the-art approaches to

*Access to This Page Needs a Subscription*

*Access to This Page Needs a Subscription*

*Access to This Page Needs a Subscription*

*Access to This Page Needs a Subscription*

model's exceptional performance, as demonstrated by its high accuracy in both seizure detection and prediction, signifies its potential as a valuable tool in epilepsy management. Moreover, its versatility makes it suitable for diverse biomedical signal analysis tasks.

Induction Motor Using Unscented Kalman Filter," in 2021 13th Iranian Conference on Electrical Engineering and Computer Science (ICEESC), 2021.

## REFERENCES

- [1] Alickovic E, Kevric J, Subasi A. Performance evaluation of empirical mode decomposition, discrete wavelet transform, and wavelet packed decomposition for automated epileptic seizure detection and prediction. *Biomedical signal processing and control*. 2018 Jan 1;39:94-102.
- [2] Slimen IB, Boubchir L, Mbarki Z, Seddik H. EEG epileptic seizure detection and classification based on dual-tree complex wavelet transform and machine learning algorithms. *Journal of biomedical research*. 2020 May;34(3):151.
- [3] A. Najari, F. Shabani, and M. Hosseynzadeh, "INTEGRATED INTELLIGENT CONTROL SYSTEM DESIGN TO IMPROVE VEHICLE ROTATIONAL STABILITY USING ACTIVE DIFFERENTIAL," *Acta Technica Corviniensis-Bulletin of Engineering*, vol. 14, no. 1, pp. 79-82, 2021.
- [4] M. Cheraghifard, G. Taghizadeh, M. Akbarfahimi, A. M. Eakman, S.-H. Hosseini, and A. Azad, "Psychometric properties of Meaningful Activity Participation Assessment (MAPA) in chronic stroke survivors," *Topics in Stroke Rehabilitation*, vol. 28, no. 6, pp. 422-431, 2021.
- [5] Carvalho VR, Moraes MF, Braga AP, Mendes EM. Evaluating five different adaptive decomposition methods for EEG signal seizure detection and classification. *Biomedical Signal Processing and Control*. 2020 Sep 1;62:102073.
- [6] Boonyakitanont P, Lek-Uthai A, Chomtho K, Songsiri J. A review of feature extraction and performance evaluation in epileptic seizure detection using EEG. *Biomedical Signal Processing and Control*. 2020 Mar 1;57:101702.
- [7] Tzimourta KD, Tzallas AT, Giannakeas N, Astrakas LG, Tsalikakis DG, Angelidis P, Tsipouras MG. A robust methodology for classification of epileptic seizures in EEG signals. *Health and Technology*. 2019 Mar 15;9:135-42.
- [8] Karabiber Cura O, Kocaaslan Atli S, Türe HS, Akan A. Epileptic seizure classifications using empirical mode decomposition and its derivative. *Biomedical engineering online*. 2020 Dec;19:1-22.
- [9] Chakraborty J, Nandy A. Discrete wavelet transform based data representation in deep neural network for gait abnormality detection. *Biomedical Signal Processing and Control*. 2020 Sep 1;62:102076.
- [10] Mamli S, Kalbkhani H. Gray-level co-occurrence matrix of Fourier synchro-squeezed transform for epileptic seizure detection. *Biocybernetics and Biomedical Engineering*. 2019 Jan 1;39(1):87-99.
- [11] George ST, Subathra MS, Sairamya NJ, Susmitha L, Premkumar MJ. Classification of epileptic EEG signals using PSO based artificial neural network and tunable-Q wavelet transform. *Biocybernetics and Biomedical Engineering*. 2020 Apr 1;40(2):709-28.
- [12] Radman M, Moradi M, Chaibakhsh A, Kordestani M, Saif M. Multi-feature fusion approach for epileptic seizure detection from EEG signals. *IEEE sensors journal*. 2020 Sep 23;21(3):3533-43.
- [13] Martis RJ, Acharya UR, Tan JH, Petznick A, Yanti R, Chua CK, Ng EK, Tong L. Application of empirical mode decomposition (EMD) for automated detection of epilepsy using EEG signals. *International journal of neural systems*. 2012 Dec 13;22(06):1250027.
- [14] Jia J, Goparaju B, Song J, Zhang R, Westover MB. Automated identification of epileptic seizures in EEG signals based on phase space representation and statistical features in the CEEMD domain. *Biomedical Signal Processing and Control*. 2017 Sep 1;38:148-57.
- [15] Aydemir E, Tuncer T, Dogan S. A Tunable-Q wavelet transform and quadruple symmetric pattern based EEG signal classification method. *Medical hypotheses*. 2020 Jan 1;134:109519.
- [16] S. Izadi, K. Jabari, M. Izadi, B. Khadem Hamedani, and A. Ghaffari, "Identification and Diagnosis of Dynamic and Static Misalignment in