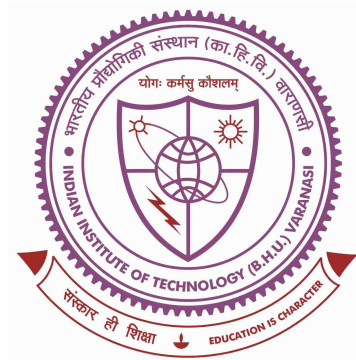


INTELLIGENT COMPUTING TECHNIQUES FOR EPILEPTIC SEIZURE PREDICTION



Thesis submitted in partial fulfillment
for the Award of Degree

Doctor of Philosophy

by

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2023

To
My Beloved Family
and
My Beloved Family-in-law

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“No one who achieves success does so without acknowledging the help of others. The wise and confident acknowledge this help with gratitude.” -Alfred North Whitehead

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ANVITI PANDEY

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List of Abbreviations

1D	One Dimensional
2D	Two Dimensional
3D	Three Dimensional
AE	AutoEncoder
AC	Artifact Combination
AD	Alzhiemers' Disease
AdaBoost	Adaptive Boosting
ADHD	Attention-Deficit/Hyperactivity Disorder
AiP	Artifacts in Pairs
AM-FBC	Adaptive Median Feature Baseline Correction
ANN	Artificial Neural Network
AR	Artifact Removal
BCI	Brain-Computer Interface
BLSTM	Bidirectional LSTM
BPNN	Backpropogation NN
BRO	Battle-Royale Optimization
BRRO	Battle Royale Search and Rescue Optimization
BS-LSTM	Bidirectional Stochastic LSTM
BSCN	Bidirectional SCN
BSO	Brainstorm Optimization
BSS	Blind Source Separation
CA	Cardiac Artifacts
CCR	Common Component Rejection
CE-stSENet	Channel-Embedding Spectral-Temporal Squeeze-and-Excitation Network

CESP	Convolutional ES Predictor
CNN	Convolutional Neural Network
Conv1Ds	1D Convolution Layer
CoTA	Combination of Three Artifacts
CV	Cross Validation
DCGAN	Deep Convolutional GAN
DFA	Detrended Fluctuation Analysis
DL	Deep Learning
DNN	Deep NN
DWT	Discrete Wavelet Transform
ECG	Electrocardiography
EDF	European Data Format
EEG	Electroencephalogram/Electroencephalography
EMD	Empirical Mode Decomposition
EMG	Electromyography
EMRVFLN	Error-Minimised RVFL Network
EOG	Electrooculography
ES	Epileptic Seizure
ESA	EEG Signal Analysis
ESN	Echo State Networks
EWT	Empirical Wavelet Transform
FB-SARO	Forensic based Search and Rescue Optimization
FBIO	Forensic Based Investigation Optimization
FC	Fully-Connected
FD	Fractal Dimension
FFT	Fast Fourier Transform
FIR	Finite Impulse Response
FLN	Functional Link Neural Network
FN	False Negative
FP	False Positives
FPA	Flower Pollination Algorithm
FPGA	Field Programmable Gates Array

FPR	False Prediction Rate
FR	Fletcher Reeves
FS	Failed Search
FV	Fisher vector
GA	Genetic Algorithm
GAN	Generative Adversarial Networks
GLCM	Gray-Level Co-occurrence Matrix
HC	Healthy Controls
HE	Hurst Exponent
HFD	Higuchi Fractal Dimension
HNN	Hierarchical Neural Network
HP	Hjorth Parameters
HT	Hilbert Transform
ICA	Independent Component Analysis
IDE	Integrated Development Environmet
iEEG	Intracranial EEG
IMFs	Intrinsic Mode Functions
KNN	K-Nearest Neighbor
LGCN	Linear Graph Convolution Network
LMD	Local Mean Decomposition
LoG	Laplacian of Gaussian
LR	Linear Regression
LSTM	Long Short-Term Memory
MA	Muscle Artifacts
MAS	Mean Amplitude Spectrum
MFS	Maximum FS
MI-EEG	Motor Imagery EEG
ML	Machine Learning
MLP	Multilevel Perceptron
MMNN	Multi-Module Neural Network
MoA	Motion Artifacts
MOFPA	Multi-Objective FPA

MPSD	Mean Power Spectral Density
MPSO	Modified PSO
MRBF	Multiscale Radial Basis Function
MSE	Mean Squared Error
NN	Neural Network
OA	Ocular Artifacts
OCNN	Optimised CNN
OLS	Orthogonal Least Squares
PCA	Principal Component Analysis
PF	Product Functions
PL	PowerLine Interference
PRD	Percentage Root Mean Square Difference
PS	Patient-Specific
PSD	Power Spectral Density
PSO	Particle Swarm Optimisation
RAM	Random Access Memory
RF	Random Forest
RGB	Red, Green and Blue
RNN	Recurrent Neural Networks
ROC	Receiver Operating Characteristic
RS-DA	Random Selection and Data Augmentation
RVFL	Random Vector Functional Link
SA	Single Artifact
SARO	Search and Rescue Optimization
SC-EEG	Single-Channel EEG
SCN	Stochastic Configuration Network
SCNN	Stacked CNN
SD	Standard Deviation
SE	Squeeze-and-Excitation
sEEG	Scalp EEG
SER	Sparsity based EEG Reconstruction
SLSTM	Stacked LSTM

SNR	Signal to Noise ratio
SOP	Seizure Occurrence Period
SPH	Seizure Prediction Horizon
SRNN	Stacked RNN
SSD	Sparse Spectrotemporal Decomposition
STFT	Short-Time Fourier Transform
SVD	Singular Value Decomposition
SVM	Support Vector Machine
TL	Team Leader
TN	True Negatives
TP	True Positives
TQWT	Tunable-Q-WT
TrP	Training Percentage
TSIN	Temporal–Spatial Information Network
UPSO	Uniform Search PSO
VMD	Variational Mode Decomposition
WDESN	Wide Deep ESN
WFB	Wavelets Filter Bank
WPFs	Wavelet Packet Features
WPT	Wavelet Packet Transform
WT	Wavelet Transform

List of Symbols

$a(\cdot)$	Activation function
b	Biases of the hidden layer
w_k^1	Central frequency
C	Contaminated signal
α	Contamination coefficient
(X_i, Y_i)	Data sample
D	Denoised signal
$Y \times Z$	Dimension of the weight values
c	Element of the signal C
r_n	Error computed from the current network
ε	Error Threshold
E	Evidence matrix
u_k^1	First modes
Q	Fixed number of nodes
Hz	Hertz
hr	Hour
λ	Input weights
imf_i	Intrinsic mode function
ϕ	Lagrangian multipliers
$L1$	Layer in LSTM
N_{max}	Maximum number of nodes that can be added to the network
$r(t)$	Monotonic trend
l	Node order
n	Number of nodes in the hidden layer of the current network