



"Classification of EEG Using Wavelet Based -Neural Networks"

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Introduction

Epilepsy is a common brain disorder that affects about 1% of the world population, where 25% of such patients cannot be treated properly by any available therapy [ENG97].

Epileptic seizures:

A sudden abnormal function of the body Loss of consciousness Increase muscular activity Abnormal sensation



Figure 1. Electroencephalogram (EEG)

Electroencephalogram (EEG) is a technique non-invasive, is clean, cost effective and safe technique for monitoring brain activity.

Research objective

This research is focused on a study of connectionist models to analyze, detect and classify to identify stages of Epilepsy using EEG.

This work aims to develop new structures of classifiers based on Wavelet Neural Networks to enhance the classification of the EEG signals.

General block diagram



Preprocessing

Some physiological researchers consider that EEG frequencies above 60 Hz are noise and can be neglected [MIR11].



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Feature extraction

In this work, wavelet analysis was used to decompose the EEG signals into delta (δ), theta (θ), alpha (α), beta (β), and gamma (γ) sub-bands.



Proposed model: Multidimensional Radial Wavelons – Feed-Forward Wavelet Neural Networks (MRW-FFWNN)



Experimental Data EEG

EEG Database from University of Bonn [EEG12] (it contains five datasets)



Each set holds 100 segments of EEG signals of 23.6 seconds. The sampling frequency of these signals was 173.6 Hz, so each segment contains 4,096 samples. Sets Z, F and S were used only for the results reported here.

EEGs states

Healthy

Inter-ictal

Ictal



Results

We present the results of a model based on wavelet analysis and neural networks for identification of seizures events of epilepsy (Ictal, Interictal and Healthy). We tested several filters, wavelets and wavelet transformations and we have used several classifiers to compare the results obtained.

- 1. FFWNN (Feed Forward Wavelet Neural Network)
- 2. MRW-FFWNN (Multidimensional Radial Wavelons FFWNN)*
- 3. SRWNN (Self-Recurrent Wavelet Neural Network)
- 4. MRW-SRWNN (Multidimensional Radial Wavelons-SRWNN)

*Proposed model

Results of classifiers based on WNN (Decision tree: Ictal-Interictal-Healthy)

Best combinations:

Feature Extraction	FFWNN (Accuracy %)					*MRW_FFWNN (Accuracy %)				
	Ictal	Inter	Healthy	Indeter	Total	Ictal	Inter	Healthy	Indeter	Total
Cheby2-SWT-Db2	90.59	46.67	71.56	12.33	72.11	92.35	40.00	72.81	13.54	71.44
Ellip-DWT-Db4	73.00	75.00	30.36	14.89	60.44	85.00	79.06	52.50	12.05	72.78

Feature Extraction	:	SRWN	IN (Acc	uracy %	5)	MRW_SRWNN (Accuracy %)				
	lctal	Inter	Healthy	Indeter	Total	Ictal	Inter	Healthy	Indeter	Total
Ellip-DWT-Db2 LeastSquares-DWT-Db1	90.94	34.81	64.84	14.52	65.11	85.94	64.81	50.97	14.21	67.56
	92.22	33.03	37.67	15.65	52.33	94.07	49.70	79.00	12.05	72.78

Characteristics of WNN:

Wavelet: Mexican hat; Iter: 100; Learning rate: 0.1; 10 executions, 40 neurons

Conclusions

♦ We present the results of a model based on MRW-FFWNN as classifier and we tested several filters, wavelets and wavelet transformations in order to find a suitable combination to improve the results reported in the classification of EEG signals.

Thanks