

Complexity analysis and dynamic characteristics of EEG using MODWT based entropies for identification of seizure onset

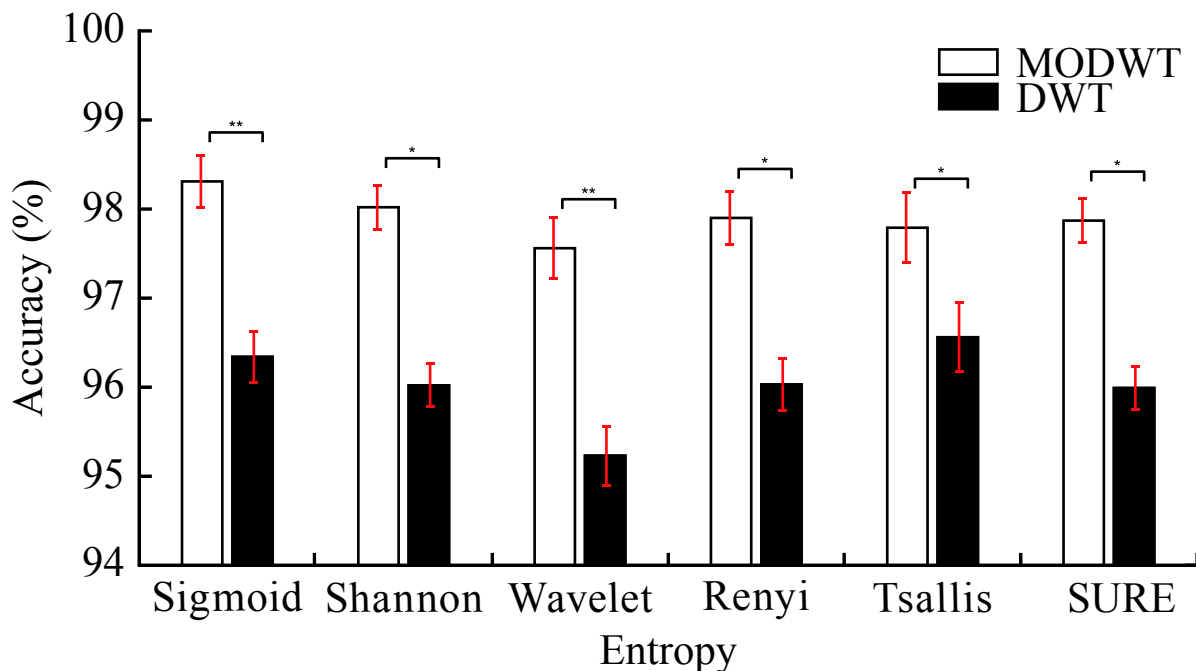
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Supplementary Fig. 1 Performance comparison between entropies derived from MODWT and DWT. * $P < 0.05$; ** $P < 0.01$. MODWT: maximal overlap discrete wavelet transform; DWT: discrete wavelet transform; SURE: Stein's unbiased risk estimator.

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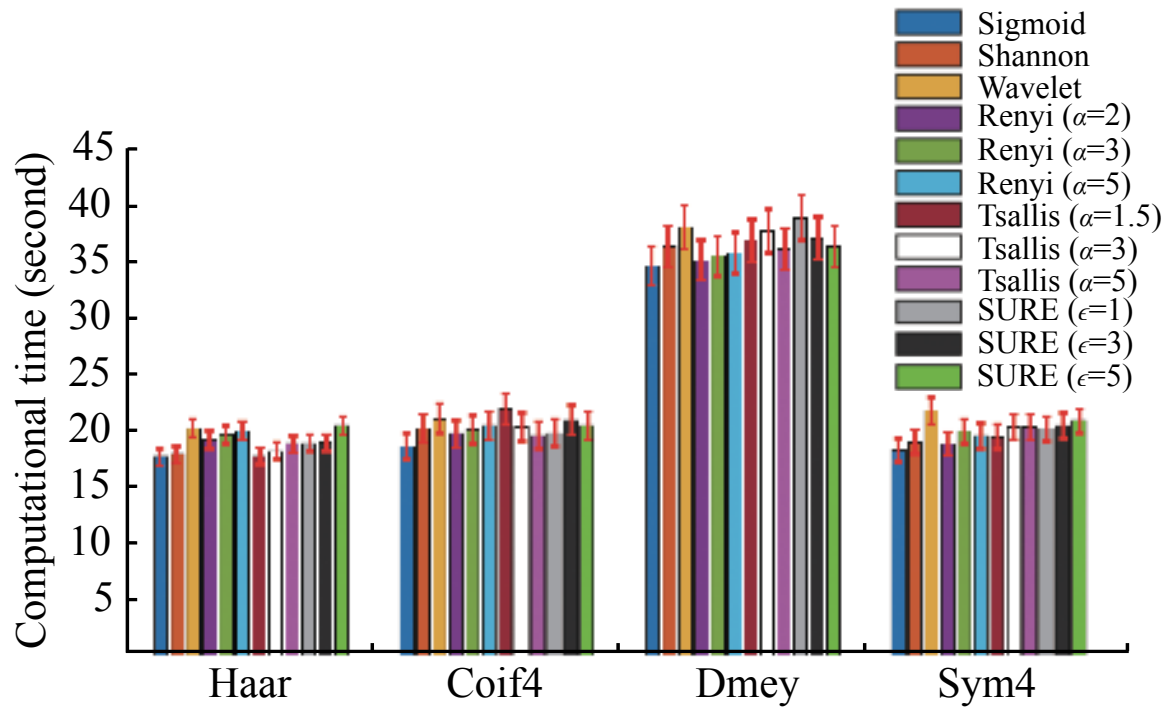
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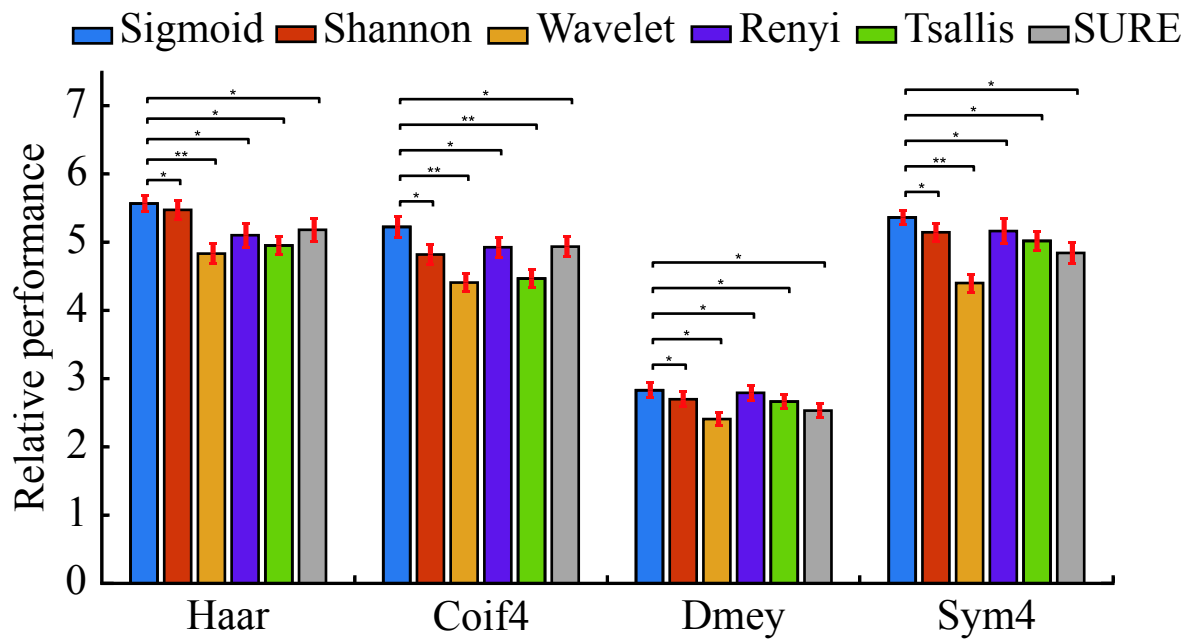
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Supplementary Table 1 Performance comparison of epileptic seizure detection methods

Authors	EEG database	Duration of EEG (hour)	No. of subjects	No. of channels	No. of seizures	Feature extraction	Classifier	Results (%)
Li <i>et al</i> ^[23]	UBonn	1.29	5	NS*	100 seizure files	MODWT+LND scale parameter and shape parameter	Random forest	Accuracy=100
Ouyang <i>et al</i> ^[24]	Epilepsy Center of the University Hospital of Freiburg, Germany	509	421	18	161	MODWT+nonlinear similarity index	Threshold	Accuracy=91.9
Liu <i>et al</i> ^[31]	Epilepsy Center of the University Hospital of Freiburg, Germany	509	421	18	161	DWT+relative energy, fluctuation index, coefficient of variation and relative amplitude	SVM	Accuracy=94.86
Vidyaratne <i>et al</i> ^[47]	CHB-MIT	884	23	23	182	Harmonic wavelet packet transform fractal dimension, spatial and temporal features	Relevance vector Machine	Sensitivity=96
	UBonn	1.29	5	NS*	100 seizure files			Accuracy=99.8
Saab <i>et al</i> ^[48]	Epilepsy Telemetry Unit at the Montreal Neurological Institute and Hospital	652	28	24	126	DWT+energy, amplitude and variance	Bayes network	Sensitivity=78
Kuhlmann <i>et al</i> ^[49]	Compumedics, Melbourne, Australia	525	21	16	88	DWT+energy, amplitude, variance, cross-correlation, and relative derivative	Bayes network	Sensitivity=79
Zandi <i>et al</i> ^[32]	Vancouver General Hospital, Germany	76	14	15	63	DWT+combined seizure index	Adaptive thresholding	Sensitivity=90.5
Hopfengärtner <i>et al</i> ^[50]	Epilepsy Center University of Erlangen, Germany	22 278	159	19	794	Moving window STFT, averaged and integrated power	Adaptive thresholding	Sensitivity=89.9
Proposed	RMCH, Bengaluru, India	58	115	19	162	MODWT+sigmoid entropy	SVM	Accuracy=98.44
	UBonn	1.29	5	NS*	100 seizure files	MODWT+sigmoid entropy	SVM	Accuracy=100
	CHB-MIT	884	23	23	182	MODWT+sigmoid entropy	SVM	Accuracy=94.51



Supplementary Fig. 2 Computational time between different entropies using MODWT for RMCH database. SURE: Stein's unbiased risk estimator; MODWT: maximal overlap discrete wavelet transform; RMCH: Ramaiah Medical College and Hospitals.



Supplementary Fig. 3 Relative performance obtained for best results among four wavelets and entropies. * $P<0.05$; ** $P<0.01$. SURE: Stein's unbiased risk estimator.