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disusun oleh: Khairul Anam dkk

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Preface

This volume is part of the three-volume proceedings of the 21st International Conference on Neural Information Processing (ICONIP 2014), which was held in Kuching, Malaysia, during November 3–6, 2014. The ICONIP is an annual conference of the Asia Pacific Neural Network Assembly (APNNA). This series of ICONIP conferences has been held annually since 1994 in Seoul and has become one of the leading international conferences in the area of neural networks.

ICONIP 2014 received a total of 375 submissions by scholars from 47 countries/regions across six continents. Based on a rigorous peer-review process where each submission was evaluated by at least two qualified reviewers, a total of 231 high-quality papers were selected for publication in the reputable series of Lecture Notes in Computer Science (LNCS). The selected papers cover major topics of theoretical research, empirical study, and applications of neural information processing research. ICONIP 2014 also featured a pre-conference event, namely, the Cybersecurity Data Mining Competition and Workshop (CDMC 2014) which was held in Kuala Lumpur. Nine papers from CDMC 2014 were selected for a Special Session of the conference proceedings.

In addition to the contributed papers, the ICONIP 2014 technical program included a keynote speech by Shun-Ichi Amari (RIKEN Brain Science Institute, Japan), two plenary speeches by Jacek Zurada (University of Louisville, USA) and Jürgen Schmidhuber (Istituto Dalle Molle di Studi sull'Intelligenza Artificiale, Switzerland). This conference also featured seven invited speakers, i.e., Akira Hirose (The University of Tokyo, Japan), Nikola Kasabov (Auckland University of Technology, New Zealand), Soo-Young Lee (KAIST, Korea), Derong Liu (Chinese Academy of Sciences, China; University of Illinois, USA), Kay Chen Tan (National University of Singapore), Jun Wang (The Chinese University of Hong Kong), and Zhi-Hua Zhou (Nanjing University, China).

We would like to sincerely thank Honorary Chair Shun-ichi Amari, Mohd Amin Jalaludin, the members of the Advisory Committee, the APNNA Governing Board for their guidance, the members of the Organizing Committee for all their great efforts and time in organizing such an event. We would also like to take this opportunity to express our deepest gratitude to all the technical committee members for their professional review that guaranteed high quality papers.

We would also like to thank Springer for publishing the proceedings in the prestigious LNCS series. Finally, we would like to thank all the speakers, authors,

and participants for their contribution and support in making ICONIP 2014 a successful event.

November 2014

Chu Kiong Loo Keem Siah Yap Kok Wai Wong Andrew Teoh Kaizhu Huang



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Table of Contents – Part I

Cogn	itive	Science
Cogn	IIIIVE	DCIGITCE

Transfer Entropy and Information Flow Patterns in Functional Brain Networks during Cognitive Activity	1
Human Implicit Intent Discrimination Using EEG and Eye Movement	11
Towards Establishing Relationships between Human Arousal Level and Motion Mass	19
Estimating Nonlinear Spatiotemporal Membrane Dynamics in Active Dendrites	27
Inter Subject Correlation of Brain Activity during Visuo-Motor Sequence Learning	35
An Agent Response System Based on Mirror Neuron and Theory of Mind	42
Dynamic of Nitric Oxide Diffusion in Volume Transmission: Model and Validation	50
A Computational Model of the Relation between Regulation of Negative Emotions and Mood	59
Neural Networks and Learning Systems - Theory and Design	
Low-Cost Representation for Restricted Boltzmann Machines	69

XIV ignate of Pontents Part Dry Universitas Jember

Neocognitron	78
Posterior Distribution Learning (PDL): A Novel Supervised Learning Framework	86
Computational Model of Neocortical Learning Process: Prototype Jing Xian Teo and Henry Lee Seldon	95
Active Learning with Maximum Density and Minimum Redundancy Yingjie Gu, Zhong Jin, and Steve C. Chiu	103
One-to-Many Association Ability of Chaotic Quaternionic Multidirectional Associative Memory Takumi Okutsu and Yuko Osana	111
An Entropy-Guided Adaptive Co-construction Method of State and Action Spaces in Reinforcement Learning	119
A Nodes Reduction Procedure for RBFNDDA through Histogram Pey Yun Goh, Shing Chiang Tan, and Wooi Ping Cheah	127
Toroidal Approximate Identity Neural Networks Are Universal Approximators Saeed Panahian Fard and Zarita Zainuddin	135
Self-organizing Neural Grove	143
Transfer Learning Using the Online FMM Model	151
A Supervised Methodology to Measure the Variables Contribution to a Clustering	159
Coupling between Spatial Consistency of Neural Firing and Local Field Potential Coherence: A Computational Study	167
Fading Channel Prediction Based on Self-optimizing Neural Networks	175
Invariant Multiparameter Sensitivity of Oscillator Networks	183

Digital Repository Univaled Contents Part Ibexv

Spatial-Temporal Saliency Feature Extraction for Robust Mean-Shift Tracker	191
BOOSTRON: Boosting Based Perceptron Learning	199
G-Stream: Growing Neural Gas over Data Stream	207
Combining Active Learning and Semi-supervised Learning Using Local and Global Consistency	215
Complex-Valued Neural Networks – Recent Progress and Future Directions (Invited Paper)	223
A Cascade System of Simple Dynamic Binary Neural Networks and Its Sparsification	231
A Model of V4 Neurons Based on Sparse Coding	239
A Fast and Memory-Efficient Hierarchical Graph Clustering Algorithm	247
Hopfield-Type Associative Memory with Sparse Modular Networks Gouhei Tanaka, Toshiyuki Yamane, Daiju Nakano, Ryosho Nakane, and Yasunao Katayama	255
Concept Drift Detection Based on Anomaly Analysis	263
Online Learning for Faulty RBF Networks with the Concurrent Fault Wai Yan Wan, Chi-Sing Leung, Zi-Fa Han, and Ruibin Feng	271
The Performance of the Stochastic DNN-kWTA Network	279
Modularity Maximization Adjusted by Neural Networks	287
A Dynamic Pruning Strategy for Incremental Learning on a Budget Yusuke Kondo and Koichiro Yamauchi	295

XVI I Grable of Contents Part Dry Universitas Jember

Neural Computing with Concurrent Synchrony	304
A Line-Partitioned Heteroassociative Memory for Storing Binary Fresnel Hologram	312
A Unified Framework for Privacy Preserving Data Clustering $\dots \dots Wenye\ Li$	319
Spiking Neural Network with Lateral Inhibition for Reward-Based Associative Learning	32'
Fuzzy Signature Neural Networks for Classification: Optimising the Structure	33
Self-organizing Map-Based Probabilistic Associative Memory Yuko Osana	342
Neural Networks and Learning Systems - Applications	
A Causal Model for Disease Pathway Discovery	350
Enhanced Non-linear Features for On-line Handwriting Recognition Using Deep Learning	358
Recognizing Human Actions by Using the Evolving Remote Supervised Method of Spiking Neural Networks	360
A Neural Networks Committee for the Contextual Bandit Problem Robin Allesiardo, Raphaël Féraud, and Djallel Bouneffouf	37
Multi-step Predictions of Landslide Displacements Based on Echo State Network	38
Discrete-Time Nonlinear Generalized Policy Iteration for Optimal Control Using Neural Networks	389

Digital Repository Univable of Contents Part 16 XVII

ANFIS-Based Model for Improved Paraphrase Rating Prediction El-Sayed M. El-Alfy	397
Contextual Bandit for Active Learning: Active Thompson Sampling Djallel Bouneffouf, Romain Laroche, Tanguy Urvoy, Raphael Feraud, and Robin Allesiardo	405
Choosing the Best Auto-Encoder-Based Bagging Classifier: An Empirical Study	413
Classification of fMRI Data in the NeuCube Evolving Spiking Neural Network Architecture	421
A Hybrid Approach to Pixel Data Mining	429
A Novel SOH Prediction Framework for the Lithium-ion Battery Using Echo State Network	438
Significance of Non-edge Priors in Gene Regulatory Network Reconstruction	446
Robust Lane Detection Based on Convolutional Neural Network and Random Sample Consensus	454
Learning Local Receptive Fields in Deep Belief Networks for Visual Feature Detection	462
Adaptive Wavelet Extreme Learning Machine (AW-ELM) for Index Finger Recognition Using Two-Channel Electromyography	471
Text Categorization Using an Automatically Generated Labelled Dataset: An Evaluation Study	479
Online Recommender System Based on Social Network Regularization	487
A Nonlinear Cross-Site Transfer Learning Approach for Recommender Systems	495

xVIII Grable of Roments Part Dry Universitas Jember

Deep Learning of Multifractal Attributes from Motor Imagery Induced EEG Junhua Li and Andrzej Cichocki	F
· ·	
A Fast Neural-Dynamical Approach to Scale-Invariant Object Detection	F
Improving Quantization Quality in Brain-Inspired Self-organization for Non-stationary Data Spaces	Ę
Utilizing High-Dimensional Neural Networks for Pseudo- orthogonalization of Memory Patterns	Ę
Adaptive Noise Schedule for Denoising Autoencoder	٤
Modeling Bi-directional Tree Contexts by Generative Transductions Davide Bacciu, Alessio Micheli, and Alessandro Sperduti	F
A Novel Architecture for Capturing Discrete Sequences Using Self-organizing Maps	Ę
An Improved Gbest Guided Artificial Bee Colony (IGGABC) Algorithm for Classification and Prediction Tasks	Ę
Artifact Removal from EEG Using a Multi-objective Independent Component Analysis Model	Ę
Cooperative Feature Level Data Fusion for Authentication Using Neural Networks	Ę
Fuzzy Output Error as the Performance Function for Training Artificial Neural Networks to Predict Reading Comprehension from Eye Gaze Leana Copeland, Tom Gedeon, and Sumudu Mendis	Ę
Part-Based Tracking with Appearance Learning and Structural Constrains	Ę

Digital Repository Univable of Contents Part Ibexix

Estimation of Hidden Markov Chains by a Neural Network	602
Corporate Leaders Analytics and Network System (CLANS): Constructing and Mining Social Networks among Corporations and Business Elites in China	610
Characteristics and Potential Developments of Multiple-MLP Ensemble Re-RX Algorithm	619
Author Index	629



Table of Contents - Part II

Kernel and Statistical Methods

A New Ensemble Clustering Method Based on Dempster-Shafer Evidence Theory and Gaussian Mixture Modeling Yi Wu, Xiabi Liu, and Lunhao Guo	1
Extraction of Dimension Reduced Features from Empirical Kernel Vector	9
Method of Evolving Non-stationary Multiple Kernel Learning	17
A Kernel Method to Extract Common Features Based on Mutual Information	26
Properties of Text-Prompted Multistep Speaker Verification Using Gibbs-Distribution-Based Extended Bayesian Inference for Rejecting Unregistered Speakers	35
Non-monotonic Feature Selection for Regression	44
Non-negative Matrix Factorization with Schatten p-norms Reguralization	52
A New Energy Model for the Hidden Markov Random Fields Jérémie Sublime, Antoine Cornuéjols, and Younès Bennani	60
Online Nonlinear Granger Causality Detection by Quantized Kernel Least Mean Square	68
A Computational Model of Anti-Bayesian Sensory Integration in the Size-Weight Illusion	76
Unsupervised Dimensionality Reduction for Gaussian Mixture Model Xi Yang, Kaizhu Huang, and Rui Zhang	84

XXII Grable of Contents Part II ry Universitas Jember

Graph Kernels Exploiting Weisfeiler-Lehman Graph Isomorphism Test Extensions	93
Giovanni Da San Martino, Nicolò Navarin, and Alessandro Sperduti	50
Texture Analysis Based Automated Decision Support System for Classification of Skin Cancer Using SA-SVM	101
In-attention State Monitoring for a Driver Based on Head Pose and Eye Blinking Detection Using One Class Support Vector Machine	110
An Improved Separating Hyperplane Method with Application to Embedded Intelligent Devices	118
Fine-Grained Air Quality Monitoring Based on Gaussian Process	
Regression	126
Retrieval of Experiments by Efficient Comparison of Marginal	
Likelihoods	135
Evolutionary Computation and Hybrid Intelligent Systems	
A New Approach of Diversity Enhanced Particle Swarm Optimization with Neighborhood Search and Adaptive Mutation	143
Data Clustering Based on Particle Swarm Optimization with Neighborhood Search and Cauchy Mutation	151
Accuracy Improvement of Localization and Mapping of ICP-SLAM via Competitive Associative Nets and Leave-One-Out Cross-Validation Shuichi Kurogi, Yoichiro Yamashita, Hikaru Yoshikawa, and Kotaro Hirayama	160
Saliency Level Set Evolution	170
Application of Cuckoo Search for Design Optimization of Heat	
Exchangers	178

Digital Repository Univade of Contents Jeann Dearth Dearth

A Hybrid Method to Improve the Reduction of Ballistocardiogram Artifact from EEG Data	186
VLGAAC: Variable Length Genetic Algorithm Based Alternative Clustering	194
Social Book Search with Pseudo-Relevance Feedback	203
A Random Key Genetic Algorithm for Live Migration of Multiple Virtual Machines in Data Centers	212
Collaboration of the Radial Basis ART and PSO in Multi-Solution Problems of the Hnon Map Fumiaki Tokunaga, Takumi Sato, and Toshimichi Saito	221
Reconstructing Gene Regulatory Network with Enhanced Particle Swarm Optimization	229
Neural Network Training by Hybrid Accelerated Cuckoo Particle Swarm Optimization Algorithm	237
An Accelerated Particle Swarm Optimization Based Levenberg Marquardt Back Propagation Algorithm	245
Fission-and-Recombination Particle Swarm Optimizers for Search of Multiple Solutions	254
Fast Generalized Fuzzy C-means Using Particle Swarm Optimization for Image Segmentation	263
Evolutionary Learning and Stability of Mixed-Rule Cellular Automata	271

XXIV Grable of Contents Part III Y Universitas Jember

Pattern Recognition Techniques	
Radical-Enhanced Chinese Character Embedding	279
Conditional Multidimensional Parameter Identification with Asymmetric Correlated Losses of Estimation Errors	287
Short Text Hashing Improved by Integrating Topic Features and Tags	295
Synthetic Test Data Generation for Hierarchical Graph Clustering Methods	303
Optimal Landmark Selection for Nyström Approximation	311
Privacy Preserving Clustering: A k -Means Type Extension	319
Stream Quantiles via Maximal Entropy Histograms	327
A Unified Framework for Thermal Face Recognition	335
Geometric Feature-Based Facial Emotion Recognition Using Two-Stage Fuzzy Reasoning Model	344
Human Activity Recognition by Matching Curve Shapes	352
Sentiment Analysis of Chinese Microblogs Based on Layered Features $Dong fang\ Wang\ and\ Fang\ Li$	361
Feature Group Weighting and Topological Biclustering Tugdual Sarazin, Mustapha Lebbah, Hanane Azzag, and Amine Chaibi	369
A Label Completion Approach to Crowd Approximation	377
Multi-label Linear Discriminant Analysis with Locality Consistency	386

Yuzhang Yuan, Kang Zhao, and Hongtao Lu

Digital Repository Univale a contents Jean Dexy

Hashing for Financial Credit Risk Analysis	395
MAP Inference with MRF by Graduated Non-Convexity and Concavity Procedure	404
Two-Phase Approach to Link Prediction	413
Properties of Direct Multi-Step Ahead Prediction of Chaotic Time Series and Out-of-Bag Estimate for Model Selection	421
Multi-document Summarization Based on Sentence Clustering	429
An Ontology-Based Approach to Query Suggestion Diversification Hai-Tao Zheng, Jie Zhao, Yi-Chi Zhang, Yong Jiang, and Shu-Tao Xia	437
Sensor Drift Compensation Using Fuzzy Interference System and Sparse-Grid Quadrature Filter in Blood Glucose Control	445
Webpage Segmentation Using Ontology and Word Matching Huey Jing Toh and Jer Lang Hong	454
Continuity of Discrete-Time Fuzzy Systems	462
Sib-Based Survival Selection Technique for Protein Structure Prediction in 3D-FCC Lattice Model	470
Tensor Completion Based on Structural Information	479
Document Versioning Using Feature Space Distances	487

xxvigrable of Contents Part 11 ry Universitas Jember

Separation and Classification of Crackles and Bronchial Breath Sounds from Normal Breath Sounds Using Gaussian Mixture Model	495
Combined Features for Face Recognition in Surveillance Conditions Khaled Assaleh, Tamer Shanableh, and Kamal Abuqaaud	503
Sparse Coding on Multiple Manifold Data	515
Mutual Information Estimation with Random Forests	524
Out-Of-Vocabulary Words Recognition Based on Conditional Random Field in Electronic Commerce	532
Least Angle Regression in Orthogonal Case	540
Evaluation Protocol of Early Classifiers over Multiple Data Sets	548
Exploiting Level-Wise Category Links for Semantic Relatedness Computing	556
Characteristic Prediction of a Varistor in Over-Voltage Protection Application	565
Optimizing Complex Building Renovation Process with Fuzzy Signature State Machines	573
News Title Classification with Support from Auxiliary Long Texts Yuanxin Ouyang, Yao Huangfu, Hao Sheng, and Zhang Xiong	581
Modelling Mediator Intervention in Joint Decision Making Processes Involving Mutual Empathic Understanding	589
Author Index	597

Table of Contents – Part III

a		-	т.	
Signal	and	Image	Proce	essing

Real Time Crowd Counting with Human Detection and Human Tracking	1
A New Method for Removing Random-Valued Impulse Noise	9
CTR Prediction for DSP with Improved Cube Factorization Model from Historical Bidding Log	17
Image Retrieval Using a Novel Color Similarity Measurement and Neural Networks	25
Bottom-Up Visual Saliency Using Binary Spectrum of Walsh-Hadamard Transform	33
Sonification for EEG Frequency Spectrum and EEG-Based Emotion Features	42
Forecasting Crowd State in Video by an Improved Lattice Boltzmann Model	50
Properties of Multiobjective Robust Controller Using Difference Signals and Multiple Competitive Associative Nets in Control of Linear Systems Weicheng Huang, Yuki Ishiguma, and Shuichi Kurogi	58
Calibrating Independent Component Analysis with Laplacian Reference for Real-Time EEG Artifact Removal	68
Unsupervised Segmentation Using Cluster Ensembles	76

XXVIII Table of Contents Part III Y Universitas Jember

Similar-Video Retrieval via Learned Exemplars and Time-Warped Alignment	85
Teruki Horie, Masafumi Moriwaki, Ryota Yokote, Shota Ninomiya, Akihiro Shikano, and Yasuo Matsuyama	
Automatic Image Annotation Exploiting Textual and Visual Saliency Yun Gu, Haoyang Xue, Jie Yang, and Zhenhong Jia	95
Classification of Fish Ectoparasite Genus Gyrodactylus SEM Images Using ASM and Complex Network Model	103
Linked Tucker2 Decomposition for Flexible Multi-block Data Analysis	111
Tatsuya Yokota and Andrzej Cichocki	111
Celebrity Face Image Retrieval Using Multiple Features Jie Jin and Liqing Zhang	119
A Novel Adaptive Shrinkage Threshold on Shearlet Transform for Image Denoising	127
Perception of Symmetry in Natural Images: A Cortical Representation of Shape	135
Image Denoising with Rectified Linear Units	142
Shape Preserving RGB-D Depth Map Restoration	150
Online Detection of Concept Drift in Visual Tracking Yichen Liu and Yue Zhou	159
Temporally Regularized Filters for Common Spatial Patterns by Preserving Locally Linear Structure of EEG Trials	167
Interactive Color Correction of Display by Dichromatic User Hiroki Takagi, Hiroaki Kudo, Tetsuya Matsumoto, Yoshinori Takeuchi, and Noboru Ohnishi	175
A Neural Ensemble Approach for Segmentation and Classification	1.04
of Road Images	185

Digital Repository Unitable of Contents Jeannibaxia

Using Biologically-Inspired Visual Features to Model the Restorative Potential of Scenes	194
Classification of Stroke Patients' Motor Imagery EEG with Autoencoders in BCI-FES Rehabilitation Training System	202
Real-Time Patch-Based Tracking with Occlusion Handling	210
Blood Cell Image Retrieval System Using Color, Shape and Bag of Words	218
Analysis of OCT Images for Detection of Choroidal Neovascularization in Retinal Pigment Epithelial Layer	226
Online Object Tracking Based on Depth Image with Sparse Coding Shan-Chun Shen, Wei-Long Zheng, and Bao-Liang Lu	234
Real-Time Compressive Tracking with a Particle Filter Framework Xuan Yao and Yue Zhou	242
Image Super-Resolution with Fast Approximate Convolutional Sparse Coding	250
Sparse Coding for Improved Signal-to-Noise Ratio in MRI	258
Scalable Video Coding Using Hybrid DCT/Wavelets Architectures	266
Image Enhancement Using Geometric Mean Filter and Gamma Correction for WCE Images	276
Autoencoder-Based Collaborative Filtering	284
Extended Laplacian Sparse Coding for Image Categorization	292

xxxigiatie of Contents Pairt mry Universitas Jember

The 2014 Cybersecurity Data Mining Competition and Workshop (CDMC2014)	
Stochastic Decision Making in Learning Classifier Systems through a Natural Policy Gradient Method	300
Quantum Inspired Evolutionary Algorithm by Representing Candidate Solution as Normal Distribution	308
Text Categorization with Diversity Random Forests	317
Unknown Attack Detection by Multistage One-Class SVM Focusing on Communication Interval	325
Analysis and Configuration of Boundary Difference Calculations Simon Dacey, Lei Song, Lei Zhu, and Shaoning Pang	333
Morphological Associative Memory Employing a Split Store Method Hakaru Tamukoh, Kensuke Koga, Hideaki Harada, and Takashi Morie	341
A Novel Hybrid Approach for Combining Deep and Traditional Neural Networks	349
A Classification Method of Darknet Traffic for Advanced Security Monitoring and Response	357
Detecting Malicious Spam Mails: An Online Machine Learning Approach	365
Intelligent Systems for Supporting Decision-Making Processes: Theories and Applications	
Freshness-Aware Thompson Sampling	373
Condition Monitoring of Broken Rotor Bars Using a Hybrid FMM-GA Model	381
Manjeevan Seera, Chee Peng Lim, and Chu Kiong Loo	

Digital Repository Unitable of contents Part III bxxxII

Employing Genetic Algorithm to Construct Epigenetic Tree-Based Features for Enhancer Region Prediction	390
Model and Algorithm for Multi-follower Tri-level Hierarchical Decision-Making	398
A Fuzzy ART-Based Approach for Estimation of High Performance Concrete Mix Proportion	407
A New Application of an Evolving Tree to Failure Mode and Effect Analysis Methodology	415
An Application of Fuzzy Adaptive Resonance Theory to Engineering Education	423
Augmented Query Strategies for Active Learning in Stream Data Mining	431
Feature Selection and Mass Classification Using Particle Swarm Optimization and Support Vector Machine Man To Wong, Xiangjian He, Wei-Chang Yeh, Zaidah Ibrahim, and Yuk Ying Chung	439
Strategic Decision Support in Waste Management Systems by State Reduction in FCM Models	447
An ELM Based Multi Agent Systems Using Certified Belief in Strength	458
Constrained-Optimization-Based Bayesian Posterior Probability Extreme Learning Machine for Pattern Classification	466

XXXII Table of Contents Part III Y Universitas Jember

Neuroeng	rineering	and	Neural	lcomputing	<u>o</u>
- 10 GI O CII	,	and	I TO CAL CA		_

Adaptive Translational Cueing Motion Algorithm Using Fuzzy Based Tilt Coordination	474
Adaptive Washout Algorithm Based Fuzzy Tuning for Improving Human Perception	483
Neurophysiology of Insects Using Microelectrode Arrays: Current Trends and Future Prospects	493
Neuron's Spikes Noise Level Classification Using Hidden Markov Models	501
Improved Robust Kalman Filtering for Uncertain Systems with Missing Measurements	509
Motor Imagery Data Classification for BCI Application Using Wavelet Packet Feature Extraction	519
Adaptive-Multi-Reference Least Means Squares Filter	527
sEMG-Based Single-Joint Active Training with <i>i</i> Leg—A Horizontal Exoskeleton for Lower Limb Rehabilitation	535
Cognitive Robotics	
Find Rooms for Improvement: Towards Semi-automatic Labeling of Occupancy Grid Maps	543

Digital Repository Unitable of contents Jeanni XXXIII

Frank Bahrmann, Sven Hellbach, Sabrina Keil, and Hans-Joachim Böhme	553
Towards Real-World Neurorobotics: Integrated Neuromorphic Visual Attention	563
Samantha V. Adams, Alexander D. Rast, Cameron Patterson, Francesco Galluppi, Kevin Brohan, José-Antonio Pérez-Carrasco, Thomas Wennekers, Steve Furber, and Angelo Cangelosi	303
On the Role of Working Memory in Trading-Off Skills and Situation Awareness in Sudoku	571
GA-Tetris Bot: Evolving a Better Tetris Gameplay Using Adaptive Evaluation Scheme	579
An Effectiveness of Model-Based Development with User Model in Consideration of Human	587
Dynamic Programming for Guided Gene Transfer in Bacterial Memetic Algorithm	596
Topological Gaussian Adaptive Resonance Associative Memory with Fuzzy Motion Planning for Place Navigation	604
Security in Signal Processing and Machine Learning	
Anomaly Based Intrusion Detection through Temporal Classification Shih Yin Ooi, Shing Chiang Tan, and Wooi Ping Cheah	612
Threshold Visual Secret Sharing Based on Boolean Operations and Random Grids	620
Wavelet Based SDA for Face Recognition	628
Essential Visual Cryptographic Scheme with Different Importance of	000
Shares	636

xxxiviable of Contents Part mry Universitas Jember

Improved Biohashing Method Based on Most Intensive Histogram	
Block Location	644
2.5D Face Recognition under Tensor Manifold Metrics Lee-Ying Chong, Andrew Beng Jin Teoh, Thian-Song Ong, and Siew-Chin Chong	653
Learning Systems for Social Network and Web Mining	
Predicting Mobile Subscriber's Behaviour from Contextual Information	
Extraction: SMS Data	661
Discovering Plain-Text-Described Services Based on Ontology	
Learning	673
A Fuzzy VSM-Based Approach for Semantic Service Retrieval Supannada Chotipant, Farookh Khadeer Hussain, Hai Dong, and Omar Khadeer Hussain	682
Maintaining Trust in Cloud Computing through SLA Monitoring Walayat Hussain, Farookh Khadeer Hussain, and Omar Khadeer Hussain	690
Author Index	699

Adaptive Wavelet Extreme Learning Machine (AW-ELM) for Index Finger Recognition Using Two-Channel Electromyography

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Abstract. This paper proposes a new structure of wavelet extreme learning machine i.e. an adaptive wavelet extreme learning machine (AW-ELM) for finger motion recognition using only two EMG channels. The adaptation mechanism is performed by adjusting the wavelet shape based on the input information. The performance of the proposed method is compared to ELM using wavelet (W-ELM0 and sigmoid (Sig-ELM) activation function. The experimental results demonstrate that the proposed AW-ELM performs better than W-ELM and Sig-ELM.

Keywords: Wavelet extreme learning machine, adaptive.

1 Introduction

A wavelet neural network (WNN) is a special case of a feed-forward neural network which its activation function is wavelets [1]. A standard gradient descend can be used to train the weight of WNN. However, drawbacks of the gradient descent method such as long training time and easy trapped to local minima have hampered the implementation of WNN in the real-time application [2]. On the other hand, an extreme learning machine (ELM) was introduced to train a single-hidden layer feed-forward networks (SLFNs) resulting in a system which is fast and able to avoid a local minima [3]. Inevitably, WNN can be constructed using SLFNs.

The combination of ELM and WNN can be conducted by simply replacing the activation function of ELM with wavelets [4] [5]. This is the simplest unification of both networks as has been done in [5]. Cao et al. [6] introduced a new combination of these two algorithms by proposing a composite function of WNN with ELM. In this method, they implemented two activation functions, a wavelet function and any piecewise function which are done in order.

Another new unification of ELM and WNN was proposed by Javed et al. [7] who proposed a summation wavelet extreme learning machine (SW-ELM). Same as Cao, Javed et al. utilized two activation functions but employed them in different ways.

These two activation functions were done in parallel and their outputs were averaged to be the output of the hidden nodes.

This paper proposes an adaptive wavelet extreme learning machine (AW-ELM), a new unification of ELM and WNN. According to WNN structure, the proposed system utilizes a wavelet function as the activation function in the hidden node. However, the activation functions are not fixed but they are adjusted regarding to the changing in the input. The sigmoid function is used to process the input information and produce translation parameters of the wavelets in the related hidden-node. In this paper, the performance of AW-ELM will be tested to classify the finger motions from the surface Electromyography signal (EMG) extracted from two-channel sources on the forearm. In addition, its classification performance will be compared with two types of ELM, ELM with wavelet activation function (W-ELM) and sigmoid activation function (Sig-ELM).

The organization of the paper is as follows: section 2 describes the theory of W-ELM and AW-ELM, and the implementation of AW-ELM for finger motion classification. Then section 3 and 4 presents the results and the discussion. Finally section 4 will conclude this paper.

2 Methods

2.1 Wavelet Extreme Learning Machine (W-ELM)

W-ELM can be considered as a special case of extreme learning machine which its activation function is wavelets. The output function of W-ELM for arbitrary samples $(x_k,t_k) \in \mathbb{R}^n \times \mathbb{R}^0$ with M hidden nodes is

$$f_i^k(\mathbf{X}) = \sum_{j=1}^{M} V_{ij} \psi_{a_j b j}(w_j, c_j, \mathbf{X}_k) \qquad i = 1, 2, ..., O$$
 (1)

where

$$\psi_{a_jb_j}(x) = \frac{1}{\sqrt{a_j}} \psi\left(\frac{x - b_j}{a_j}\right), \qquad j = 1, 2, \dots, M$$
(2)

in which a_j and b_j are dilatation and translation parameters of the wavelets, respectively. An initialization of dilatation and translation parameters, a_j and b_j , in WNN is an important issue. The initialization should consider the input information in order to let the time domain of the wavelet covering the input domain. According to [1], suppose the input vector x_k has the domain $[x_{kmin}, x_{kmax}]$, t^* and σ^* are the centre and the radius of the mother wavelet $\psi_{a_ib_i}$, then domain of $\psi_{a_ib_i}$ is given by:

$$[b_i + a_i(t^* - \sigma^*), b_i + a_i(t^* + \sigma^*)]$$

Meanwhile, the input information range for *i*th hidden layer can be calculated as:

 $\left[\sum_{i=1}^{N} w_{ji} x_{i \min} , \sum_{i=1}^{N} w_{ji} x_{i \max}\right]$

where w_{ji} is the weight connecting the *j*th hidden layer the *i*th input. The wavelet can cover the input space if

$$b_{j} + a_{j} (t^{*} - \sigma^{*}) = \sum_{i=1}^{N} w_{ji} x_{i \min}$$
(3)

and

$$b_{j} + a_{j} (t^{*} + \sigma^{*}) = \sum_{i=1}^{N} w_{ji} x_{i \max}$$
 (4)

From equation (9) and (10), we can calculate a_i and b_i as:

$$a_{j} = \frac{1}{2\sigma^{*}} \left(\sum_{i=1}^{N} w_{ji} x_{i \max} - \sum_{i=1}^{N} w_{ji} x_{i \min} \right)$$
 (5)

$$b_{j} = \frac{1}{2\sigma^{*}} \left(\sum_{i=1}^{N} w_{ji} x_{i \max} (\sigma^{*} - t^{*}) + \sum_{i=1}^{N} w_{ji} x_{i \min} (\sigma^{*} + t^{*}) \right)$$
(6)

2.2 Adaptive Wavelet Extreme Learning Machine (AW-ELM)

The Proposed Structure

The proposed AW-ELM is depicted by Fig. 1. If M is the number of hidden node and N is the number of input, then the input of the hidden layer P_j is given by

$$P_{j}(x) = \sum_{i=1}^{N} x_{i} w_{ji} + c_{j}$$
 $j = 1, 2, ..., M$ (7)

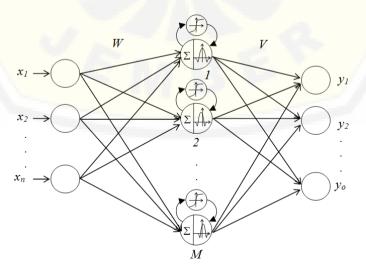


Fig. 1. The proposed adaptive wavelet extreme learning machine

where x_i are the input variables, w_{ji} are the weights of the connection between *i*th input and *j*th hidden nodes, and c_j denotes the bias of *j*th hidden layer. Using equation (8), the output of the hidden node is given by:

$$\psi_{a_j b_j}(P_j(x)) = \psi\left(\frac{P_j(x) - b_j}{a_j}\right), \qquad j = 1, 2, ..., M$$
 (8)

In this proposed work, the Mexican Hat function [6] is used as the mother wavelet $\psi_{a_ib_i}$ as described in fig. 2a, and defined as

$$\psi(x) = e^{-x^2/2} (1 - x^2) \tag{9}$$

Therefore, the wavelet activation function of AW-ELM is:

$$\psi_{a_{j}b_{j}}(P_{j}) = e^{-0.5\left(\frac{P_{j}-b_{j}}{a_{j}}\right)^{2}} \left(1 - \left(\frac{P_{j}-b_{j}}{a_{j}}\right)^{2}\right)$$
(10)

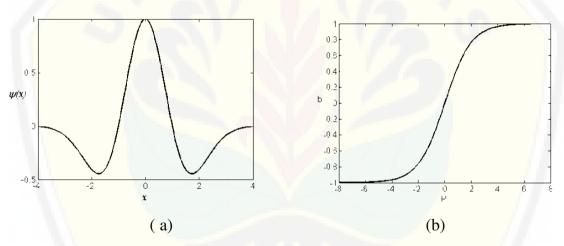


Fig. 2. Two difference functions used in this work: (a) The mother wavelet of the Mexican hat (b) A nonlinear function to produce b_i

In this proposed AW-ELM, the dilatation parameters a_j are fixed and initialized using Equation (5). As for the translation parameters b_j , they are varied according to the input information and driven by a nonlinear function f(.) as follows:

$$b_i = f(P_i) \tag{11}$$

where

$$f(P_j) = \frac{2}{1 + e^{-P_j}} - 1 \tag{12}$$

as depicted in Fig. 2b. Eventually, a new structure of an adaptive W-ELM is presented in Fig. 1. A small circle on the top of each hidden node is used to adjust the b parameters in order to change the shape of the wavelet. Thus, the output of AW-ELM is:

$$f_{i}^{k}(\mathbf{X}) = \sum_{j=1}^{M} V_{ij} \psi_{a_{j}b_{j}}(w_{j}, c_{j}, \mathbf{X}_{k}) = \sum_{j=1}^{M} V_{ij} \psi_{a_{j}b_{j}}(P_{j}(\mathbf{X}_{k})) \quad i = 1, 2, ..., O$$
(13)

The Learning Algorithm

For the desired output:

$$D = (\boldsymbol{d}_1^T \quad \boldsymbol{d}_2^T \quad \dots \quad \boldsymbol{d}_L^T)_{LxO}$$
 (14)

The AW-ELM described in (13) can be written as a linear system as follows:

$$HV = D \tag{15}$$

where

$$\mathbf{H} = \begin{bmatrix} \boldsymbol{\psi}_{a_1b_1}(P_1(\mathbf{x}_1)) & \cdots & \boldsymbol{\psi}_{a_Mb_M}(P_M(\mathbf{x}_1)) \\ \vdots & \vdots & \vdots \\ \boldsymbol{\psi}_{a_1b_1}(P_1(\mathbf{x}_L)) & \vdots & \boldsymbol{\psi}_{a_Mb_M}(P_M(\mathbf{x}_L)) \end{bmatrix}_{LxM}$$
(16)

$$\mathbf{V} = (\mathbf{v}_1^T \quad \mathbf{v}_2^T \quad \cdots \quad \mathbf{v}_O^T)_{MxO}^T \tag{17}$$

V can be obtained by solving the least-square solution of (15) and given by:

$$\hat{\mathbf{V}} = \mathbf{H}^{\dagger} \mathbf{D} \tag{18}$$

where H[†] is the Moore-Penrose generalized inverse of the matrix H.

The training algorithm of AW-ELM can be implemented as follows:

Algorithm of AW-ELM. Given a training set $\aleph = \{(x_k, t_k) \mid x_k \in \mathbb{R}^n, t_k \in \mathbb{$

- (1) Randomly assign vector matrix W and initialize the hidden node parameters a_j , j = 1,2,..., M according to (5)
- (2) Calculate the input hidden layer P_i in (7)
- (3) Calculate \underline{b}_i in (11) and (12)
- (4) Calculate the hidden layer output H in (16)
- (5) Calculate the output weight \hat{V} in (18)

2.3 AW-ELM for Finger Motion Recognition

The proposed recognition system consists of the same stages as depicted in Figure 3. Firstly, signals from two-channel EMG located on the forearm were acquired by a

data acquisition device from eight subjects. The experimental procedures for the data acquisition could be referred to [8]. Then the filtering and windowing was applied to the collected data before being extracted using a time domain (TD) and autoregressive (AR) features.

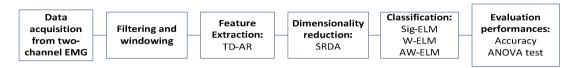


Fig. 2. The motion finger classification using AW-ELM

The features were extracted from the time domain feature set which consists of Waveform Length (WL), Slope Sign Changes (SSC), Number of Zero Crossings (ZCC), and Sample Skewness (SS). In addition, some parameters from Hjorth Time Domain Parameters (HTD) and Auto Regressive (AR) Model Parameters were included. To reduce the dimension of the features, SDRA was employed. All features were concatenated and reduced using SRDA. SRDA is an extension of LDA that can deal with singularity and a large data set. The 200 ms window length was applied to the signal to comply with the real time application along with a 25 increment.

The reduced feature set resulted in the previous stage is utilized in the classification. The objective of the classification that was performed using AW-ELM and other ELM classifiers is to recognize ten classes of the individual and combined finger movements consisting of the flexion of individuated fingers. They consisted of Thumb (T), Index (I), Middle (M), Ring (R), Little (L) and the pinching of combined Thumb–Index (T–I), Thumb–Middle (T–M), Thumb–Ring (T–R), Thumb–Little (T–L), and the hand close (HC). Finally, statistical analyses were performed to validate the result.

3 Results and Discussion

In this section, the performance of the proposed AW-ELM was compared to wavelet extreme learning machine (W-ELM) and sigmoid extreme learning machine (sig-ELM). All classifiers classified ten finger motions using EMG signal from two channel electrodes. The four-fold cross validation was used to validate the classification results. Simulation was done in the MATLAB 8.3 environment running on 2.8 GHz PC.

Table 1 shows the classification results of three classifiers in recognizing ten finger motions classes defined in 2.4. In all ELMs, the number of hidden nodes varied from 50 up to 500. The results indicate that the average accuracy the proposed of AW-ELM was higher than standard W-ELM in all cases. Likewise, the AW-ELM performance is better than Sig-ELM in all hidden node numbers except 50 and 75. In these two hidden numbers, the Sig-Elm achieved better accuracy that AW-ELM. Overall, the adaptation of wavelet shape using a sigmoid function in AW-ELM could enhance the performance of the original wavelet extreme learning machine and in several condition, could attain better performance than Sig-ELM.

Table 1. The average classification accuracy of AW-ELM across eight subjects using four-fold cross validation compared with W-ELM and Sig-ELM

# Hidden Node	Accuracy (%)			
# IIIddell Node	W-ELM	AW-ELM	Sig-ELM	
50	91.07 ± 0.17	91.57 ± 0.08	91.65 ± 0.08	
75	91.56 ± 0.14	91.93 ± 0.14	91.97 ± 0.10	
100	91.79 ± 0.10	92.05 ± 0.09	92.01 ± 0.08	
125	91.90 ± 0.08	92.08 ± 0.09	92.03 ± 0.10	
150	91.94 ± 0.11	92.06 ± 0.10	92.04 ± 0.10	
175	91.98 ± 0.09	92.06 ± 0.09	92.04 ± 0.08	
200	91.99 ± 0.08	92.04 ± 0.08	92.01 ± 0.06	
500	91.79 ± 0.08	91.56 ± 0.06	91.37 ± 0.06	

Table 2. Processing time of different ELM classifiers

#Hidden Training Time (s)			Testing Time (s)			
Node	W-ELM	AW-ELM	Sig-ELM	W-ELM	AW-ELM	Sig-ELM
50	0.16 ± 0.01	0.19 ± 0.02	0.12 ± 0.00	0.03 ± 0.00	0.06 ± 0.00	0.01 ± 0.00
75	0.28 ± 0.01	0.33 ± 0.01	0.19 ± 0.00	0.06 ± 0.01	0.07 ± 0.00	0.02 ± 0.00
100	0.38 ± 0.02	0.45 ± 0.02	0.27 ± 0.01	0.07 ± 0.00	0.10 ± 0.00	0.02 ± 0.00
125	0.59 ± 0.05	0.70 ± 0.06	0.48 ± 0.07	0.09 ± 0.00	0.14 ± 0.00	0.03 ± 0.00
150	0.71 ± 0.01	0.81 ± 0.01	0.51 ± 0.01	0.12 ± 0.00	0.19 ± 0.01	0.04 ± 0.00
175	0.93 ± 0.06	1.06 ± 0.05	0.69 ± 0.05	0.14 ± 0.00	0.22 ± 0.00	0.04 ± 0.00
200	1.07 ± 0.08	1.22 ± 0.05	0.85 ± 0.06	0.17 ± 0.00	0.26 ± 0.00	0.05 ± 0.00
500	4.20 ± 0.08	5.42 ± 0.08	2.82 ± 0.10	0.75 ± 0.01	1.28 ± 0.01	0.12 ± 0.00

In terms of processing time, the ELM using a sigmoid function (Sig-ELM) spent less training time than W-ELM and AW-ELM in as shown in Table 2. Table 2 shows that the larger the number of the hidden node, the longer the time difference between AW-ELM and Sig-ELM. Likewise, in the testing time, AW-ELM is the slowest system. The adaptive mechanism adds the processing time in both the training and testing trials.

Table 3. The p-value of anova test on the classification accuracy between AW-ELM and other tested classifiers

#Hidden	<i>p</i> -value			
Node	AW-ELM & W-ELM	AW-ELM & Sig-ELM		
50	0.0000	0.0000		
75	0.0000	0.1283		
100	0.0000	0.0006		
125	0.0000	0.0610		
150	0.0000	0.3477		
175	0.0021	0.5746		
200	0.0098	0.0552		
500	0.0000	0.0000		

The one-way ANOVA test was done to evaluate the improvement significance of AW-ELM compared to W-ELM and Sig-ELM as presented in Table 4. Table 4 shows that p-values on the comparison of AW-ELM and W-ELM are less than 0.05. In other words, the performance improvement in recognizing ten finger motions by AW-ELM is significantly achieved. Furthermore, the performance of AW-ELM and Sig-ELM in some cases is significantly different in the hidden number node 50, 100 and 500 whereas it is significantly similar in other hidden node numbers. Nevertheless, the AW-ELM produced better accuracy in most trials than Sig-ELM.

4 Conclusion

This paper proposed a novel ELM i.e. an adaptive wavelet extreme learning (AW-ELM) for recognizing finger motions using two-channel EMG signals. The adaptation mechanism of the proposed method is conducted by adjusting the shape of the wavelet based on the information provided in the input. The experimental results showed that the proposed AW-ELM improved the performance of the original wavelet ELM in all cases tested and performed better than Sig-ELM in most cases observed. In the future, the performance of AW-ELM should be compared with other well-known classifiers such as support vector machine (SVM) and Linear Discriminant Analysis (LDA).

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