



University of
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A Report on Recent Computer Science Conferences

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March 6, 2009



WorldComp'08 Overview

- 2008 World Congress in Computer Science, Computer Engineering, and Applied Computing
 - July 14 – 17, 2008, Las Vegas, NV
 - “Many of the 25 joint conferences are the premier conferences for presentation of advances in their respective fields.”
 - Over 2000 researchers from 82 different countries
 - International Conference on Artificial Intelligence (ICAI'08)
 - International Conference on Frontiers in Education: Computer Science and Computer Engineering (FECS'08)



WorldComp'08 Conferences

- **International Conference on Bioinformatics and Computational Biology (BIOCOMP'08)**
- **International Conference on Computer Design (CDES'08)**
- **International Conference on Computer Graphics and Virtual Reality (CGVR'08)**
- **International Conference on Communications in Computing (CIC'08)**
- **International Conference on Scientific Computing (CSC'08)**
- **International Conference on Data Mining (DMIN'08)**
- **International Conference on e-Learning, e-Business, Enterprise Information Systems, and e-Government (EEE'08)**
- **International Conference on Engineering of Reconfigurable Systems and Algorithms (ERSA'08)**
- **International Conference on Embedded Systems and Applications (ESA'08)**
- **International Conference on Foundations of Computer Science (FCS'08)**
- **International Conference on Frontiers in Education: Computer Science and Computer Engineering (FECS'08)**
- **International Conference on Grid Computing and Applications (GCA'08)**
- **International Conference on Genetic and Evolutionary Methods (GEM'08)**
- **International Conference on Artificial Intelligence (ICAI'08)**
- **International Conference on Internet Computing (ICOMP'08)**
- **International Conference on Wireless Networks (ICWN'08)**
- **International Conference on Information and Knowledge Engineering (IKE'08)**
- **International Conference on Image Processing, Computer Vision, and Pattern Recognition (IPCV'08)**
- **International Conference on Information Theory and Statistical Learning (ITSL'08)**
- **International Conference on Machine Learning; Models, Technologies and Applications (MLMTA'08)**
- **International Conference on Modeling, Simulation and Visualization Methods (MSV'08)**
- **International Conference on Parallel and Distributed Processing Techniques and Applications (PDPTA'08)**
- **International Conference on Security and Management (SAM'08)**
- **International Conference on Software Engineering Research and Practice (SERP'08)**
- **International Conference on Semantic Web and Web Services (SWWS'08)**



FECS'08 Overview

- Conference included recent research results and trends in CS and CE education:
 - Collaborative learning
 - Evaluation strategies
 - Game programming
 - Learning tools and emulators
 - Online learning
 - Teaching strategies
 - Virtual worlds
 - Web-based educational tools



ICAI'08 Overview

- Conference included recent research results and trends in various AI topics:
 - Agent technologies
 - Genetic algorithms
 - Fuzzy logic and applications
 - Information retrieval
 - Knowledge-based systems
 - Natural language processing
 - Neural networks and applications
 - Robotics



ICAI'08: Neural Networks

- Conference included several useful results in the area of Neural Networks (NNs).
- NNs have been successfully used for a variety of applications, ranging from stock price prediction to automatic target detection and recognition.
- Especially useful as pattern recognizers/classifiers due to their:
 - Robustness
 - Fault tolerance
 - Universal function approximation capability

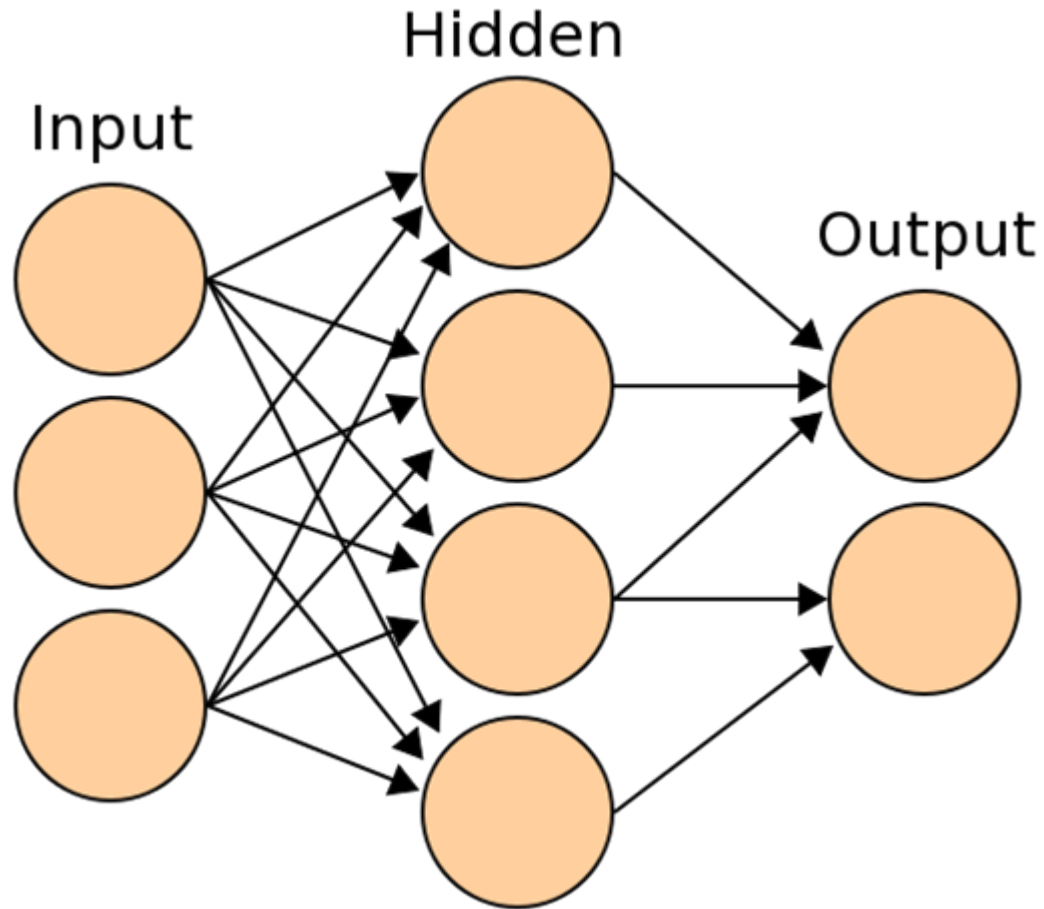


Background Information

- A multi-layer feed-forward NN with enough nodes and hidden layers can approximate any bounded function and its derivative to any arbitrary accuracy given a “large enough” training set.
- A network with N inputs, one hidden layer of H units, and a total of W weights requires $O(W/e)$ training patterns to yield error $< e$ on the test set.



Neural Network Model





Is Your Neural Network Learning or Memorizing?

Feng-Jen Yang
Computer Science Department
Prairie View A&M University

- Ideally, learning should be meaningful and comprehensive but rote learning may happen in real life as well as in simulated biological intelligence such as artificial neural networks.
- The author performed a series of experiments on five different NNs to learn the same quadratic equation:

$$y(x) = 2x^2 + 3x + 1, \text{ where } 1 \leq x \leq 30.$$

- Analyzed three statistical measures to indicate whether a NN is learning or memorizing:
 - Learning curve, mean and variance of error ratios
- All NNs had the same input layer, output layer, learning rate, etc, but different number of neurons in the hidden layer.
- Hypothesis: Do more neurons lead to better learning?



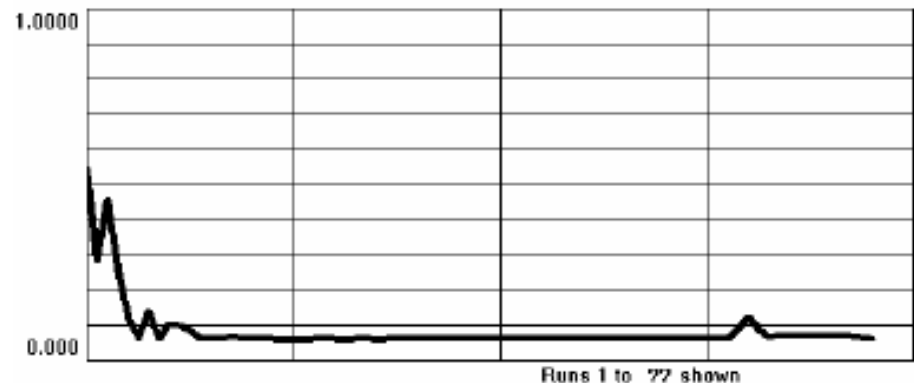
Is Your Neural Network Learning or Memorizing (cont'd)?

- Having more than enough neurons can turn a neural network from learning into memorizing and hurt the overall learning process.
- Overall comparison:

Number of Hidden neurons	10	15	100	200	300
Learning Curve	Smooth	uneven	uneven	rough	very rough
Mean of Error Rate	Small	slightly high	high	very high	very high
Variance of Error Rate	Small	slightly high	high	very high	very high
Learn Behavior	meaningful	less meaningful	memorizing	highly memorizing	extremely memorizing

- Sample test results & learning curve NN# 4 (200 hidden neurons):

X	$Y_d(x)$	$y(x)$	$e(x)$
4	45	12.16	0.73
8	153	33.78	0.78
12	323	136.87	0.58
16	561	452.11	0.19
20	861	958.33	0.11
24	1225	1391.30	0.14
28	1653	1634.30	0.01
The mean of $e(x)$: 0.36			
The variance of $e(x)$: 0.32			
The minimum of $e(x)$: 0.11			
The maximum of $e(x)$: 0.78			





A Practical Method for Finding an Efficient Static Learning Rate for an ANN

Celso Camilo¹ and Keiji Yamanaka²

¹Federal University of Grande Dourados, Brazil

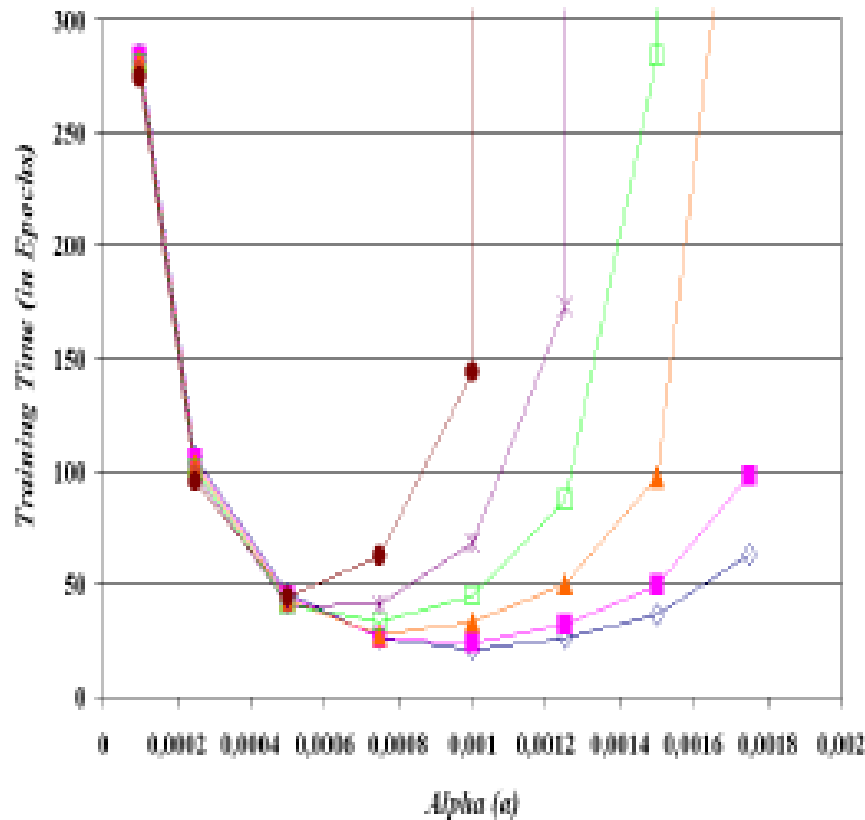
²Federal University of Uberlandia, Brazil

- Learning algorithms often have difficulties converging quickly:
 - Need to adjust learning rate carefully
- The authors propose a method for finding an efficient static learning rate (α) for single-layer nets based on:
 - Input vector dimensions
 - Input values



A Practical Method for Finding an Efficient Static Learning Rate for an ANN (cont'd)

- When the number of inputs increases, the best α decreases:

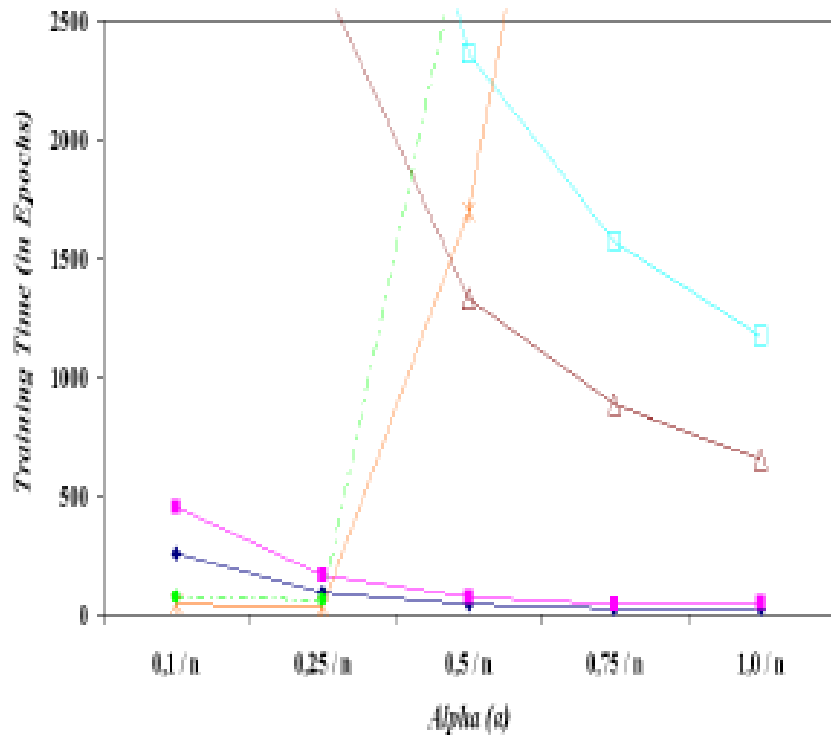


Code	Code	Inputs	Outputs
◆	1	900 (Matrix 30 x 30)	6
■	2	961 (Matrix 31 x 31)	6
▲	3	1089 (Matrix 33 x 33)	6
□	4	1225 (Matrix 35 x 35)	6
✕	5	1369 (Matrix 37 x 37)	6
●	6	1600 (Matrix 40 x 40)	6



A Practical Method for Finding an Efficient Static Learning Rate for an ANN (cont'd)

- When the input value ≤ 1 , the best α tends to the upper limit ($1/n$), and when the value > 1 , it tends to the lower limit ($0.1/n$):



Code	input neurons	input values
1	900 (Matrix 30 x 30)	(1 or -1)
2	1600 (Matrix 40 x 40)	(1 or -1)
3	900 (Matrix 30 x 30)	(0.2 or -0.2)
4	1600 (Matrix 40 x 40)	(0.2 or -0.2)
5	900 (Matrix 30 x 30)	(2.2 or -2.2)
6	1600 (Matrix 40 x 40)	(2.2 or -2.2)



Use of Neural Networks to Detect Impaired Cheating on the Computerized CERAD Word List Memory Test

Timothy Chan and Amar Raheja
Department of Computer Science
California State Polytechnic University

- The computerized CERAD Word List Memory Test is used by the Long Term Care insurance industry to screen applicants for mild cognitive impairment, and for early detection of Alzheimer's Disease.
- The authors developed a method to detect cheating by cognitively impaired individuals using a two-stage approach:
 - 1) distinguish cheaters from non-cheaters
 - 2) distinguish impaired cheaters from unimpaired-cheaters
- They employed the use of NNs and compared results with a statistical method.





Use of Neural Networks to Detect Impaired Cheating on the Computerized CERAD Word List Memory Test (cont'd)

- They found that the NN did a better job in stage 1 (2Stage1) by virtue of having a dominating average ROC (Receiver Operating Characteristic) curve.
- They also explored a one stage approach (1Stage).
- They compared the two stage and one stage NN and found that the former approach marginally outperformed the latter with a higher sensitivity yet similar positive predictive value (PPV) and specificity.

Confusion matrix		Predicted	
		Negative	Positive
Actual	Negative	5652.5±55.4	128.08±7.6
	Positive	10.3±2.2	4.71±1.4

TABLE II

CONFUSION MATRIX FOR 1Stage. PPV IS $3.56 \pm 1.05\%$, SENSITIVITY IS $31.49 \pm 8.05\%$, WHILE SPECIFICITY IS $97.78 \pm 0.13\%$.

Confusion matrix		Predicted	
		Negative	Positive
Actual	Negative	5629.7±54.7	134.3±8.5
	Positive	7.48±1.73	5.26±1.53

TABLE III

CONFUSION MATRIX FOR TWO STEP METHOD: COMBINING 2Stage1 AND 2Stage2. PPV IS $3.77 \pm 1.09\%$, SENSITIVITY IS $41.27 \pm 9.66\%$, WHILE SPECIFICITY IS $97.67 \pm 0.15\%$.

- Results still outperformed other statistical methods.



A Comparison of Neural Network Architectures for Handwritten Digit Recognition

Eman El-Sheikh, Bradley Swain, and Mohamed Khabou
University of West Florida

- Project focus:
 - Development, use, and analysis of an artificial neural network architecture for recognizing handwritten digit data.
- Project Goal:
 - To study the effects of varying architectures, learning rates, number and sensitivity of neurons on the ability to distinguish the digits.
- Potential applications:
 - Mail sorting
 - Automated check reading
 - Data entry for hand-held devices



Neural Network Architecture

- Architecture was implemented in Java, using a modular and parameterized design model.
- Network parameters include:
 - number of input neurons
 - number of hidden layers
 - number of neurons per hidden layer
 - number of output neurons
 - learning rate
 - number of training epochs



Data Description

- A collection of handwritten digits collected by U.S. Postal Service from ZIP codes.
- Three digit features were chosen and compared for their ability to distinguish the digits:
 - Information measure
 - Orthogonality measure
 - Combination measure
- All three feature data sets consisted of data for 6000 digits, of which 2000 were used for training, and 3500 used for testing.



Results

Hidden Layer(s)	Info Features	Time (mm:ss)	Orthogonal Features	Time (mm:ss)	Combination Features	Time (mm:ss)
25/20	90.63%	04:18	93.77%	04:20	93.74%	04:15
25/15	90.83%	03:56	93.34%	03:59	93.34%	03:59
25	90.54%	03:18	93.57%	03:19	93.43%	03:20
15	89.91%	02:08	92.86%	02:08	92.83%	02:08



Results

- Using simple back-propagation, feed-forward neural networks, we could correctly predict approx. 94% of 3500 handwritten digits within four minutes.
- An asymptotic bound was approached for the time/accuracy trade-off.
- An increase of only 0.2% accuracy cost another minute of processing time.



Conclusions

- Architecture successfully classified test data using each of three feature sets.
- Orthogonal features produced best results, which influenced the combination features.
- Information feature set did not perform as well due to similarity in features.
- Single hidden layer network with 25 nodes performed nearly as well as two hidden layer network with 25 & 20 nodes, with better time efficiency.