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**FINAL REPORT:**

**ONR Contract #N00014-89-J-1249**

**DESIGN AND HARDWARE IMPLEMENTATION OF NEURAL SYSTEMS**

Principal Investigator : Paul Mueller

Project Period: 10/01/88 - 9/30/91

**SUMMARY:**

This project was a three year continuation of **ONR Contract #N00014-87-K-0780**. It had two major aims. One was the design and computer simulation of specific neural circuits modeled after the early vision system of higher vertebrates. The second aim was the design and construction of electronic analog neural systems that could be used for implementing such a vision system and possibly other neural systems. Our initial approach to the hardware implementation was to build an entirely hardwired system using VLSI assemblies of electronic neurons and separate resistive synapse connection matrices fabricated by CVD deposition of amorphous silicon. During further simulation studies of the vision system it became clear that at this early state of development a completely hardwired system would be too restrictive and for this reason we felt that we needed a system that would incorporate modifiable synapses and a programmable connection architecture. Rather than restrict the design of such a system by tailoring it to the specific architectures of the vision system, we decided to design a general purpose programmable neural analog computer that would be useful not only for vision but also for other tasks of real world computation and other problems suitable for neural computation. A prototype of this machine was developed during the past three years and constructed in collaboration with Corticon INC, a small company founded by the Principle Investigators. The machine performed as specified. A software operating system was developed and a variety of simple application programs were written and run on the machine. They included the real time decomposition of acoustical patterns, recognition of simple phonemes and examples of optical pattern recognition such as OCR. A much larger machine suitable for commercial applications is currently beeing built.

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In addition to the hardware project, theoretical work was carried out on fundamentals of pattern analysis and recognition in Neural Nets, in particular on methods and strategies for pattern decomposition and representation as well as one shot learning algorithms for pattern decoding useful in speech recognition.

## **DETAILED REPORT:**

### **Simulations of a Neural Vision System**

The vision system simulations implemented in a simplified manner the basic functions of a retina and the primary visual cortex as far as they are concerned with the decomposition of static images into visual primitives such as oriented edges and lines of different spatial scales and contrast sign as well as their end points and curvature. The system consisted of many layers of neurons each layer tuned to specific primitive. Initial simulations were performed on systems that had only feed-forward connections from a receptor array to arrays of ON CENTER and OFF CENTER units and from there to orientation arrays. During further work it became clear that certain problems such as the disambiguation of contrast direction for oriented edges and improved orientation tuning required mutually inhibitory connections between units in different arrays and simulation of such connections by digital methods turned out to be extremely time consuming. Our inability to explore the synaptic gains and architectures for these inhibitory connections as well as the difficulty of determining an optimal architecture and connectivity through simulation were some of the reasons for considering a programmable analog network. Another reason came from difficulties in manufacturing precise resistive networks with available technology. Simulations at the available level of accuracy of the synaptic weights showed that results were marginal and that we would either have to spend considerable effort in trying to achieve better resistance tolerances or to develop programmable synapses that would allow us to tune each synapse individually and would provide us with means for rapidly exploring different connectivities. These reasons as well as our growing appreciation for the need of a programmable machine led us to direct our hardware efforts towards the design and implementation of a general purpose neural machine.

### **General Purpose Neural Analog Computer**

The following is a brief description of the design. The architecture of the machine is loosely based on the cortex of higher vertebrates, in the sense that there are sets of individual neurons each of which receives only a limited number of inputs, that is not every neuron is connected to every other neuron. However in contrast to biological systems, our machine would be able to modify the connection architecture by external control and thus allow exploration of different architectures in addition to adjustment of synaptic weights

and neuron parameters. Although the connections and synaptic gains are under digital control the machine runs in analog mode. A modular design allows expansion to any degree and at moderate to large size, i.e.  $10^3$  to  $10^6$  neurons, operational speed and power would exceed any currently available digital computer.

The machine contains large numbers of the following separate elements: **neurons, synapses, routing switches, and connection lines.** Arrays of these elements were fabricated on VLSI chips which are mounted on planar leadless chip carriers each of which forms a separate module. These modules are connected directly to neighboring modules. By increasing the number of modules the machine can be expanded to any degree. All parameters such as the connections between different neuron groups, individual synaptic gains and neuron parameters, (threshold and time constant) are set digitally from separate microprocessor. In addition to the analog outputs from the neurons which feed into the other neurons, the neural outputs are also multiplexed, A-D converted and fed into a digital computer that uses this information in a learning mode to set connections and synaptic weights.

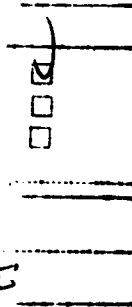
The neuron design is based on an earlier version which we have used in a neural machine for acoustical pattern recognition.

The synaptic modules contain arrays of synapses. The synaptic gain or weight of each synapse is set from the microprocessors by serial digital inputs which are stored at each synapse. Dynamic range of the gain covers the range from 0 to 10 with a logarithmic 5 bit resolution, a 6th bit determines the sign. The synapses for the larger machine will have an 8bit resolution. Implementation of the weights is by current mirrors.

The switch modules serve to route the signals between the neurons and thus determine the connectivity. Each module contains a cross-point array of analog switches which are set by serial digital input. There is also a set of serial switches that can disconnect selected input and output lines. The line modules contain fixed lines that serve as fillers between the switch modules.

All chips are mounted on identical leadless planar chip carriers. Input and output lines of the carriers are arranged at right angles, with identical leads on opposite sides. Groups of chip carriers are mounted on special boards and connected directly to each other.

Connections, synaptic gains and time constants are set by the central computer either from stored program libraries that contain connection architecture and synaptic constants appropriate for specific tasks or from connection parameters which are computed from neuron outputs on the basis of various learning algorithm. Special routing routines track already occupied lines, switches and synapses and adjust connection pathways accordingly.



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Higher level control software, allowing the automatic transfer of conceptual networks into the machine has also been developed. In this case the neurons in the conceptual network are first partitioned and grouped according to the degree to which their inputs originate from common sources. The groups are then assigned to specific modules in the physical network by a placement routine and subsequently interconnected by an autorouting program based on min-cut procedures. This program generates settings of switches and component parameters that are loaded into the neural computer.

The network is connected directly to the outside world through parallel analog I/O buffers. There are 352 analog I/O lines available. In addition to the analog I/O, each neuron chip contains an analog multiplexer that enables the digital host to monitor and store the neuron activity for graphic display and the implementation of programmed learning algorithms. The multiplexer feeds over a common line into a 250 KHz A/D board located in the host. The monitor software generates either arrays of selected time segments of each neuron output that are displayed as separate graphs, or it generates a continuous gray scale display of activity for all neurons. Neuron activity is also displayed directly by an LED Panel. The neuron activity patterns read by the host also form the basis for implementation of different learning algorithms in which the network and the host form a closed loop. The learning of a simple OCR task by backpropagation has been demonstrated.

#### **Performance of simple computational tasks.**

The machine has been programmed for several tasks in order to evaluate its performance. A few examples are discussed. In all cases the actual neural computations (not the learning) are performed in real time, with response time limited only by the bandwidth of the neurons which was maximally 300 KHz.

The first example, was a "Winner Take All" net which, depending on the inhibitory feedback gain, either extracts the largest output among 32 neurons or enhances the contrast between the different outputs. The network settles within the time constant of the neuron outputs and shows no oscillations. Simulations of this small network on a SUN 4 are slower by a factor of 100. Since this is a fully connected net the speed ratio would scale with  $n^2$  of the number of neurons.

Other small circuits, have been programmed, including an associative net of 24 neurons, a motor control circuit that models the last stage of the control of saccadic eye movements, a neural integrator and several circuits for the computation of time domain pattern primitives.

An example of a time domain operation which shows small circuits that are tuned to the frequency of amplitude modulation of a carrier frequency. Such circuits are useful in the recognition of sonar patterns. Several coupled oscillators and central pattern generators have also been programmed. By making use of threshold bias, feedback and synaptic time constants, the networks can generate arbitrary dynamic activity patterns in the absence of external inputs.

The example of character recognition demonstrated that the system can learn by backpropagation in a simple 4 layer network. Development of alternative learning algorithms for the machine is still at an early stage. Various algorithms can be run either by the host alone or with the neural computer in loop. Since the weights are computed and updated by the host, the learning speed for error correcting algorithms is improved only in those cases where the actual performance computation is the rate limiting factor such as in networks with extensive front ends or networks having lateral or feedback connections or different synaptic time constants. However, we are developing single shot learning schemes for pattern recognition where the synapse parameters are determined from the input patterns without iterative adjustments.

The machine is especially suited for the real-time analysis of dynamic patterns such as speech. As an example a more elaborate network was programmed for the primary decomposition of acoustical patterns in real time and the recognition of simple phonemes. This network, involving most of the available neurons, performs an initial decomposition of acoustical patterns. It receives input from a set of 8 bandpass filters (200 - 3000 Hz) and extracts local maxima of amplitude vs frequency, ( $d^2E/dS^2$ ), local rates of rise and fall of amplitude ( $+dE/Dt$ ), and local rise and fall of frequency i.e. motion ( $+dS/dT$ ) of activity along the frequency axis. The activity of the decomposition neurons were used as inputs to phoneme specific neurons which became active when a particular phonem was spoken to the machine.

Several conclusions can be drawn from these tests. First of all the machine performance has met all expectations regarding noise levels, accuracy, stability, programming flexibility and processing speed. Particularly the routing space proved entirely adequate. There is therefore no reason to project an increase in the ratio of switch modules to neuron and synapse modules for larger machines. No major design change for the neurons is needed. The synapse gain (weighting) scheme could be improved by increasing the resolution for middle range gains (between 0.1 and 1). The range and resolution of synaptic time constants seems adequate.

The larger machine currently under development will be capable of maximal speeds of beteen  $10^{11}$  and  $10^{12}$  connections/s.

**Publications:**

**Mueller, P.,** Computation of temporal pattern primitives in a neural net for speech recognition. **International Neural Network Society**, Boston, MA, Sept. 6-10, 1988.

**Mueller, P., Van der Spiegel, J., Blackman, D., Dao, J., Donham, C., Furman, R., Hsieh, D.P. and Loinaz, M.,** General purpose neural analog computer. **IEEE Conference on Neural Information Processing Systems**, Denver, CO, Nov. 28 - Dec. 1, 1988.

**Mueller, P. and Blackman, D.** Orientation tuning and orientation specificity in a neuron network for visual image decomposition. **AIP Conference on Neuronal Networks for Computing**, Snowbird, Utah, April, 1989.

**Mueller, P.,** Neural computation of pattern primitives, (Abstract) **AIP Conference**, Snowbird, Utah, 1988.

**Mueller, P., Hopfield, J., Lau, C. and Mead, C.** Report on a workshop "Hardware Implementation of Neuron Nets and Synapses", P. Mueller, ed., San Diego, CA, 1988.

**Mueller, P., Hopfield, J., Lau, C. and Mead, C.** Workshop on hardware implementation of neuron nets and synapses. **IEEE Circuits and Devices Magazine**, 4(3):23-24, 1988.

**Mueller, P., Blackman, D. and Furman, R.E.,** Neural Computations of Visual Images. In: **An Introduction to Neural and Electronic Networks**, S.F. Zornetzer, J. Davis and C. Lau, editors, Academic Press, 1989.

**Mueller, P. and Blackman, D.** Orientation tuning and orientation specificity in a neuron network for visual image decomposition. **AIP Conference on Neuronal Networks for Computing**, Snowbird, Utah, April, 1989.

**Mueller, P., Van der Spiegel, J., Blackman, D., Chiu, T., Clare, T., Dao, J., Donham, C., Hsieh, T.-p. and Loinaz, M.** Programmable Analog Neural Computer and Simulator. In: **Advances in Neural Information Processing Systems 1**, D.S. Touretzky, ed., page 712-719, Morgan Kaufmann Publishers, San Mateo, CA, 1989.

**Mueller, P., Van der Spiegel, J., Blackman, D., Chiu, T., Clare, T., Dao, J., Donham, C., Hsieh, T.-p. Loinaz, M.** A General Purpose Analog Neural Computer. **Proceedings of IEEE and ICNNS International Joint Conference on Neural Networks**, 1989.

**Mueller, P., Van der Spiegel, J., Blackman, D., Chiu, T., Clare, T., Dao, J., Donham, C., Hsieh, T.-p., Loinaz, M.** Design and Fabrication of VLSI Components for a General Purpose Analog Neural Computer. Analog VLSI Implementation of Neural Systems, IEEE Carver Mead, Mohammed Ismail, Eds., Kluwer Academic Publishers, Boston, MA. 1989.

**Van Der Spiegel, J., Mueller,P., Blackman, D., Donham, C., Etienne, R., Aziz, P., Choudhury, L., Jones, L., Xin, J.,** Artificial Neural Networks: Principles and VLSI Implementation. Proc. of the 5th Microelectronis Congress, Campinas, Brazil. July 1990.

**Mueller,P.,** Computation of Pattern Primitives in a Neural Network for Acoustical Pattern Recognition. Proceedings of the IJCNN 1990 Wash. I 149-151 .

**Mueller,P.,Van der Spiegel,J., Agami,V. Blackman,D. Chance,P. Donham,C., Etienne,R. Kim,J. Massa,M.,** Design and Performance of a Prototype Analog Neural Computer. Proceedings of the International Joint Conference on Neural Networks, Seattle 1991.

**Mueller, P. Van der Spiegel,J., Blackman,D. Chance,P. Donham,C.,Etienne,R. Kim,J. A** Prototype Neural Analog Computer. Proceedings of the Second Annual DOD Conference on Neural Networks. Huntsville Alabama. 1991.

**Van Der Spiegel J., Mueller P., Blackman,D. Chance,P., Donham,C., Etienne,R., Kim,J., Kinget,P.** A Multichip Analog Neural Network. Proc. Intern. Symp.VLSI Tech. Syst. Applications, VLSITSA 1991 pp 64-68.

**Van Der Spiegel, J., Mueller, P., Blackman D., Chance,P. Donham,C., Etienne ,R., Kinget.** An Analog Neural Compuer With Modular Architecture for Real-Time Dynamic Computations. IEEE Transactions On Solid State Circuits V 27 1992 (accepted).

**Mueller P., Van Der Spiegel, J. Agami,V. Blackman,D. Chance,P. Donham,C., Etienne,R. Flinn,J., Kim,J. Massa,M. Samarasekera S.,** Design of a Neural Analog Computer. Proceedings of the 2nd International Conference on Microelectronics for Neural Networks, 1991 Kryll and Method Verlag Munich, 347-357.

Paul Mueller

A handwritten signature in black ink, appearing to read 'Paul Mueller', written in a cursive style.

Principal Investigator