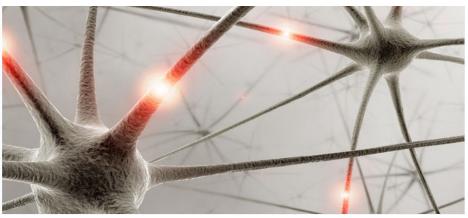


NEUROMEM, NEUROMORPHIC IC



NATIVELY PARALLEL ZISC-BASED ARCHITECTURE (ZERO INSTRUCTION SET FOR LEARNING/RECOGNITION)



Feb 2015



General Vision

- Inventor and owner of the NeuroMem CM1K
- Business Model
 - IP licensing
 - Evaluation platforms
 - Consulting Services
 - Technology Transfer Program
 - Hardware design
 - API development
 - Application demonstrators
 - Collaborations:
 - Asahi Glass(4yrs), 22 intl joint patents
 - Intel (2yrs), outcome: IP License
 - Anurag, FPGA blackbox license



NeuroStack



BrainCard (front & back)



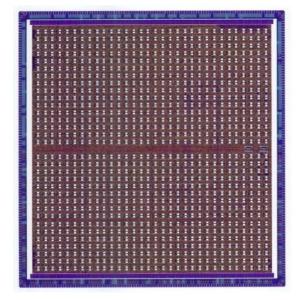


How is NeuroMem different?



NeuroMem CM1K IC

- RCE and KNN chip
- Match 1 among N in 500 ns to 2.5 μsec
- Highly scalable
- Regular architecture, just neurons
- No fetch and decode
- Patented WTA bus (no cross bar)
- Low power (<0.5 watts)
- Self trainable
- Orthogonal inter-chip connectivity
- Commercially available (IC, Source and FPGA IP)



← 1024 identical neuromorphic memories, all interconnected (intra & inter hip)



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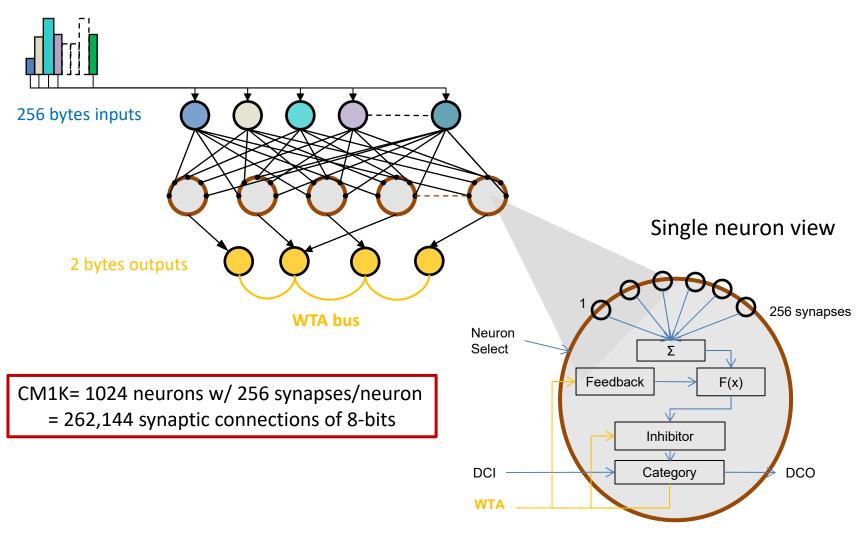
The pillars of neuromorphic model...

- Broadcast Mode: query/stimulus is broadcasted to all the neurons simultaneously
- Deterministic search time: Firing time does not increase with the scaling up of the network
- Winner takes all: Inhibit the weak responders autonomously in the same deterministic time
- Uncertain response: Sort responses autonomously in order of "lesser confidence" when multiple conflicting neurons fire
- Unknown response: Enable the dynamic addition of new knowledge
- Back propagation of error: Inhibit erroneous spiking neurons autonomously
- No fetch and decode of program instruction: Software is definitively contrary to the biological model, else it's simulation, not neuromorphic...
- Beyond biology: Fast upload download enabling knowledge proliferation (<u>some</u> <u>dream of it</u>).

All implemented in the NeuroMem CM1K IC (continued...)

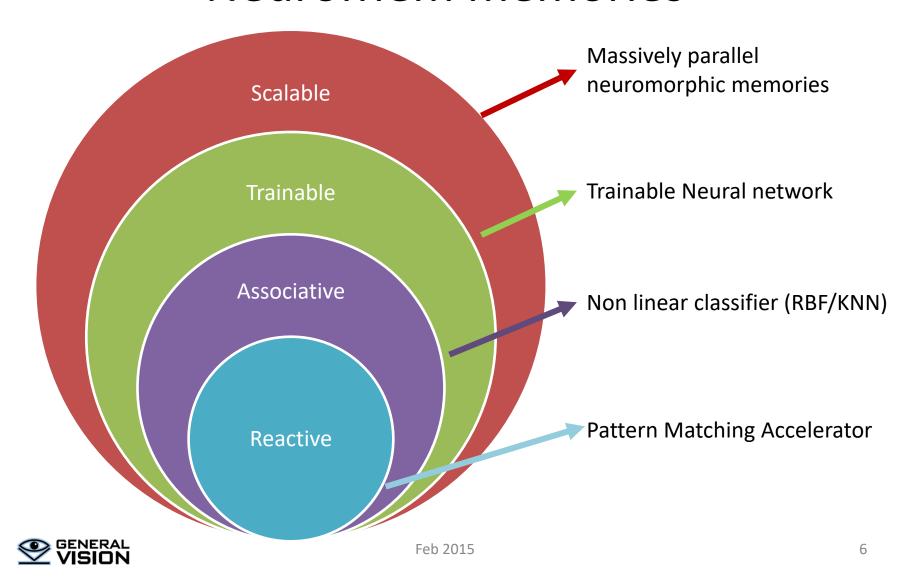


NeuroMem CM1K seen as a 3-layer NN

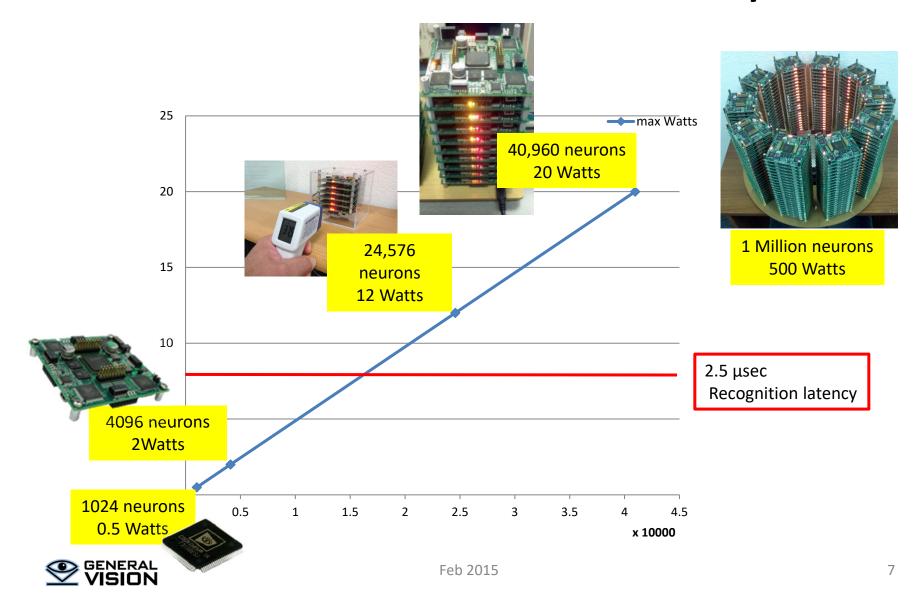




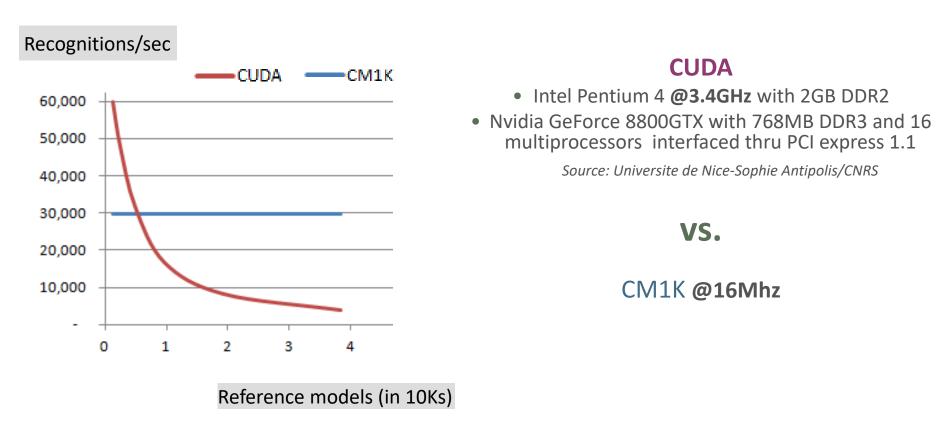
The functional facets of the NeuroMem memories



Demonstrated Scalability



K-Nearest Neighbor (KNN) Benchmarking



CONCLUSION: CM1K outperforms CUDA by an increasing factor as the number of reference models gets bigger because CUDA is highly impacted by this value when CM1k is not



Usage Models

Usage models Feature extractions Identification Data **Exact matching** Raw samples NeuroMem Text Classification **Signatures** Voice **Anomaly detection** Signal Iconic representation Fuzzy lookup table **Images** Sparse codes Videos Clustering Deep learning



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Deployed Applications with CM1K

- Deployed and non confidential
 - Fish inspection (GV, 2003)
 - Surface inspection (GV, 2009)
 - Adaptive Optics (ISL, 2011)
 - Face recognition (ANURAG, 2011)
- In progress
 - Network intrusion detection (Argonne Natl Lab)
 - Self Organizing Map (Synaptics.org)
 - Deep learning for traffic sign recognition (GV)
 - MVED (ISL)
 - CM1K as restricted Bolzman machine (ITT New Delhi)
 - IntelliGlass (GV & Asahi Glass)



Industrial off-shore fish inspection

Al Magazine Volume 29 Number I (2008) (© AAAI)

Artides

Fish Inspection System Using a Parallel Neural Network Chip and the Image Knowledge Builder Application

Anne Menendez and Guy Paillet

The fish industry is very competitive. Fleet owners are very interested in filling their boats as fast as possible with the fewest and most qualified personnel, thus reserving maximum occupancy for their refrigerated storage. During an expedition, which can last between one to two weeks, the fish processing machinery operates around the clock (figure 1). Typically, fishes are brought on the boat and dropped into metal pockets that convey them through cleaning, cutting, and filleting machines. Anomalies, which

frame grabbers, PCs, and image-processing software. None of these attempts have led to a usable offshore system because of the high nonlinearity of the problem.

A neural network approach was the only possible way to deliver a system that could be both adaptive and trainable by the fishers themselves. A hardware neural network was the best way to deliver a reliable and fast system that featured both a small footprint and affordable cost. Typical fish species to be recognized include ill-defined herrings or mackerels.

■ A generic image learning system, CogniSight, is being used for the inspection of fishes before filleting off-shore. More than 30 systems have been deployed on seven fishing vessels in Norway and Iceland over the past three years. Each CogniSight system uses four neural network chips (a total of 312 neurons) based on a natively parallel, hard-wired architecture that performs real-time learning and non-linear classification (RBF). These systems are trained by the ship crew using Image Knowledge Builder, a "show



Adaptive Optics







Quick Links



Wavefront sensing: the neural approach

Real-time wavefront measurement by embedded artificial intelligence: No computer required for operation

Current prototype features

- Commercial Micro-lens array: 150 μm pitch
- 11 x 11 micro-lens cells sampled
- Calculation of one Zernike polynomial: ~80 μs
- Resolution: 0.05 lambda (wavelength)
- · Numerical and bar-graph results display
- · Reconstructed wave front pseudo-color plot
- ISL modified <u>V1KU camera</u> from <u>General Vision</u>.



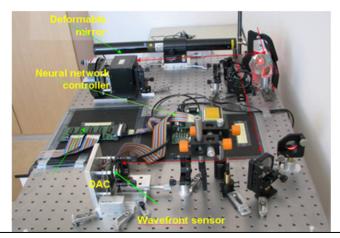
First experimental layout of a neural network-based adaptive optics controller type

Features

- 37 channel mirror
- 11 x 11 microlens array
- 8 bit DAC
- 60 fps, i.e. 18 ms / loop

Performances

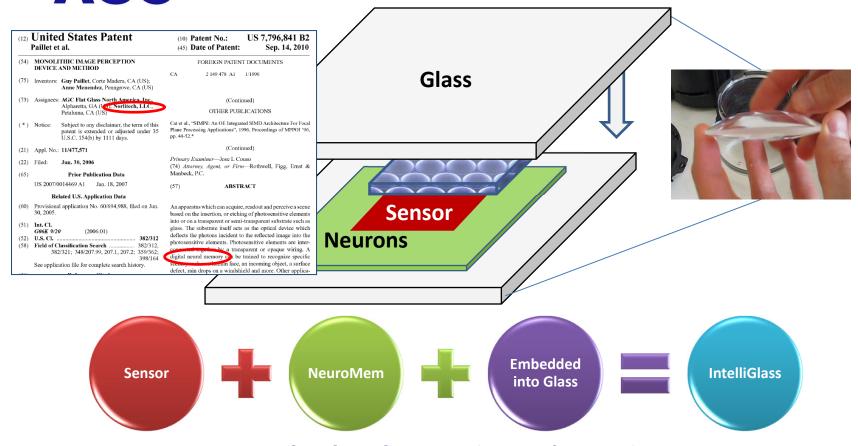
- All-electronic control time: 1.2 ms
- 830 Hz loop rate may be expected with high-speed camera interface: in development.







IntelliGlass collaboration



The glass that perceives and recognizes in consumer electronics, building automation, medical, automotive, and more.



www.general-vision.com

THANK YOU

