

Biologically sound neural networks for embedded systems using OpenCL István Fehérvári, Anita Sobe, Wilfried Elmenreich University of Klagenfurt/Lakeside Labs

Introduction

Biologically sound spiking neural networks are powerful but require high performance computing. New embedded chips with OpenCL support are perfectly suitable for this kind of tasks.





Parallel computing with OpenCL

Spiking Neural Networks

We implemented and simulated biologically plausible spiking neural networks with the following spike response model:

 $u_i(t) = \eta(t - t_i) + \sum_j w_{ij} \sum_k \varepsilon_{ij}(t - t_i) + t_j^k$

Membrane potential function

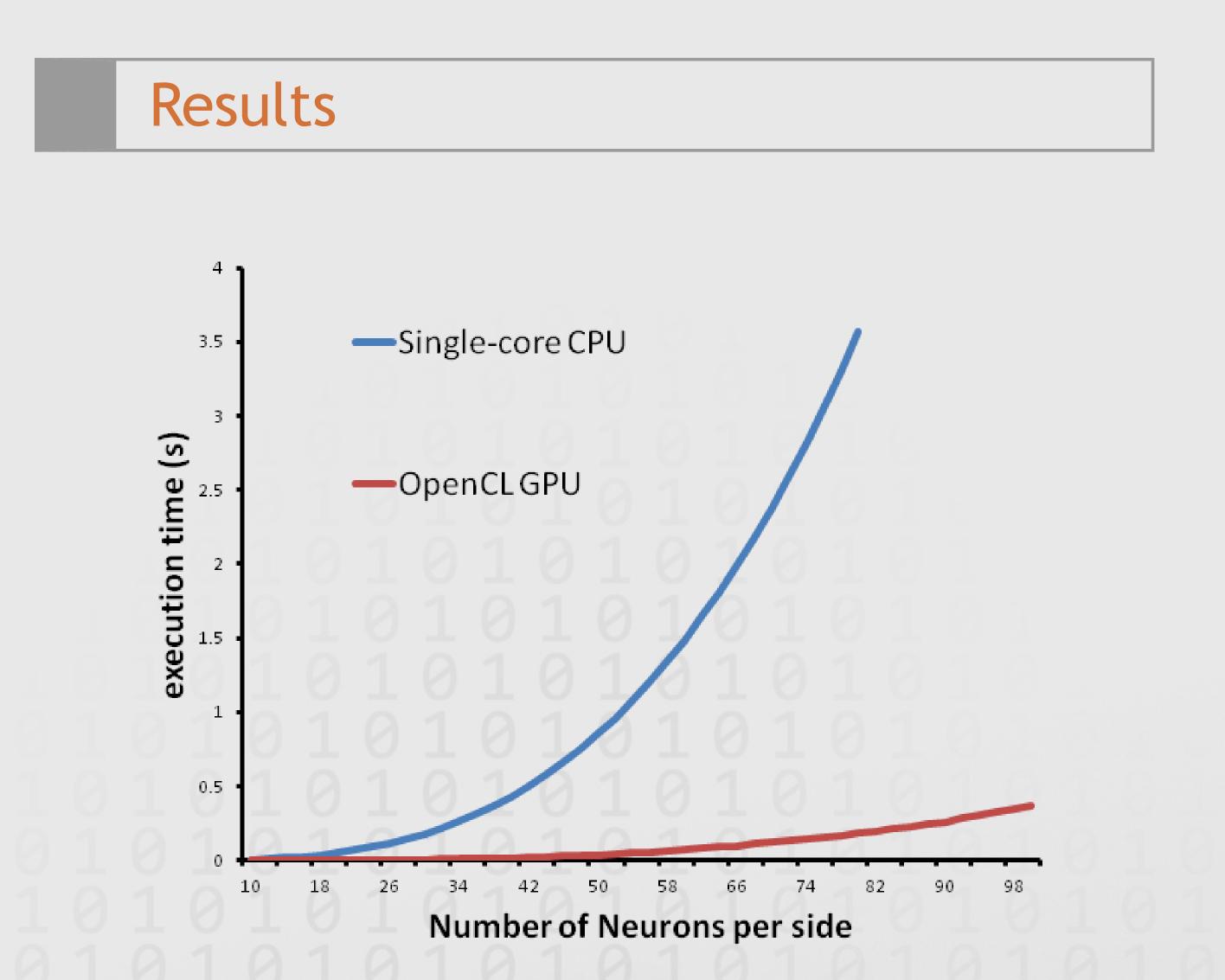
 $\varepsilon(x,y) = \frac{e^{-t/\tau_x}}{\tau_s} \int_0^x e^{-\tau(1/\tau_m - 1/\tau_s)} H(t-\tau) d\tau$

Postsynaptic potential function

 $\eta(x) = -\eta_0 e^{-x/\tau_f} H(x)$

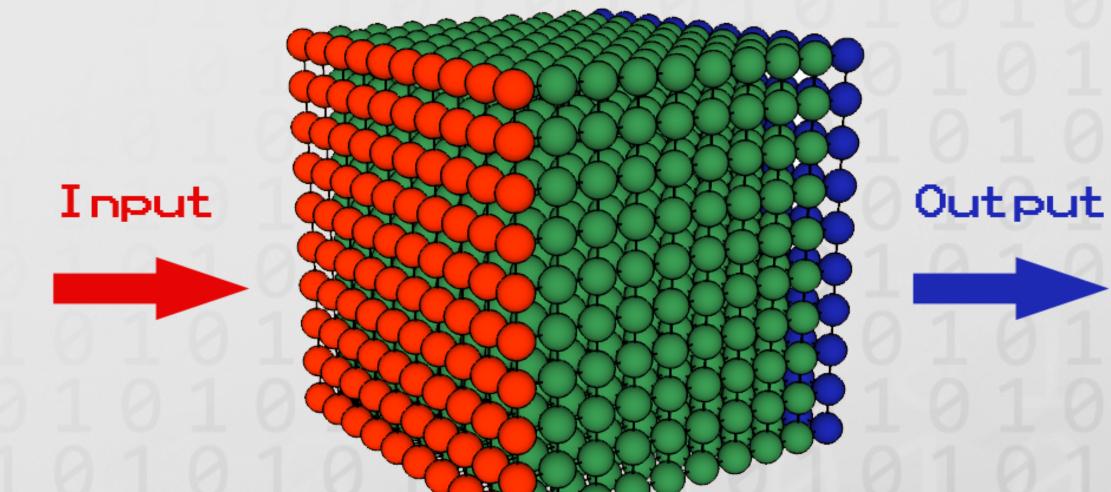
Refractory response function

- Neural networks are easily parallelizable
- OpenCL allows for implementing portable data-parallel code
- For maximum performance data transfer between host and chip has to be minimized
- Hardware's memory buffers are an upper limit for network size



Network structure

- Spatially homogeneous N × N × N network
- Neurons connect to the local Moore neighborhood (range 2)
- $5^3-1 = 124$ synapses per neuron
- Neurons on the front are used as input
- Neurons on the back are used as outputs
- Best fit for processing sequence of images



Iterated the network 1000 times, grid size N=10..100
Measured the execution time of one step of the network
Input 8 bit grayscale images

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