Large scale modeling of neuro-synaptic activity ad plasticity

Pier Stanislao Paolucci for INFN Roma APE LAB.

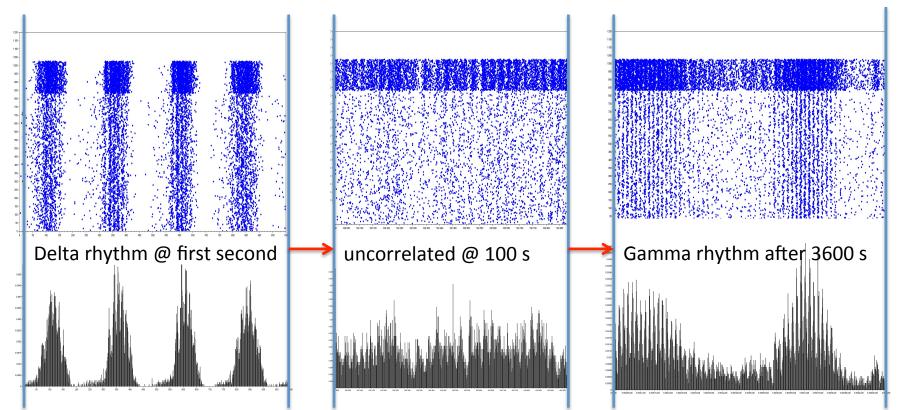


Neuro-synaptic activity and Plasticity

- Key areas of present INFN activity on large scale neural modeling
 - Coding of scalable Parallel/Distributed simulator
 - INFN developed the DPSNN-STDP simulator in the EURETILE FET Project. Proven simulation up to 6.6 G synapses, 128 cores.
 - See arXiv:1310.8478 (Apr 2014)
 - Comparison with experimental neuro-biological data and calibration of the INFN simulator
 - Will be performed in the CORTICONIC FET project (starting from Oct 2014, end Dec 2015) (cooperation with ISS, TUM, IDIBAPS)
 - Interface with experimental systems / inclusion of the simulator into robotic platforms
 - Will be investigated by the INFN "COSA" (iniziativa di gruppo 5), start Jan 2015
 - Co-design of simulation code and execution platform
 - The plan is to start from "COSA" and "CORTICONIC" to prepare the participation to a future European project on this topic



Emergent Biological Behaviour: Spontaneous Evolution of Rythmic Activity due to Polychronism and Synaptic Plasticity



As synaptic weights evolve according to STDP (synaptic spike-timing dependent plasticity, initial **delta** frequency oscillations (2-4 Hz @ first second activity) dissolves for a while into **uncorrelated** Poissonian activity (activity @ 100 seconds) and then **gamma** frequency activity emerges (30-100 HZ @ 3600 seconds)

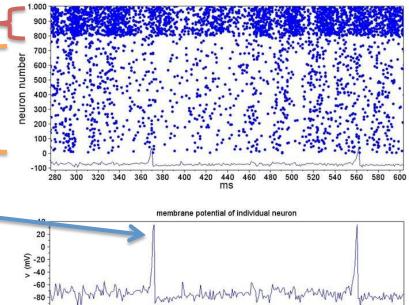


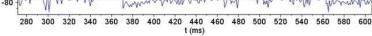
DPSNN-STDP simulates Spiking Activity and Synaptic Plasticity (already proved from 100 K up to 6.6 Giga synapses, from 1 to 128 software processes)

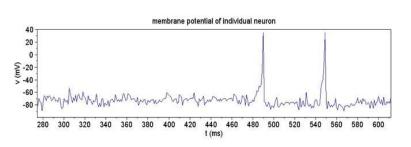
- The picture represents the evolution of a neural network computed by the DPSNN-STDP code
- This picture:

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- 200 inhibitory neurons
- 800 excitatory neurons
- total 100 000 synapses
- Time resolution: 1ms (horizontal axis)
- Each dot in the raster gram represents an individual spike
- The evolution of the membrane potential of each neuron is simulated
- The evolution of individual synaptic strength is computed (not shown in the picture)
- Polychronism: individual synaptic delays are taken into account
- Individual connections and neural types can be programmed







Collective Spiking Rastergram and activity of individual neurons

Measurements - MPI version of DPSNN-STDP

From 200 K to 6.6 Giga synapses, 1 K to 32.8 Million neurons

From 1 to 32 K cortical columns, max bi-dim grid 256x128

From 1 to 128 software processes mapped onto 2.4 GHz cores

Total synapses	200 K	800 K	3.2 M	12.8 M	51.2 M	204.8 M	819.2 M	3.2 G	6.6 G
Total neurons	1 K	4 K	16 K	64 K	256 K	1024 K	4.096 M	16.4 M	32.8 M
Grid of neural	1 x 1	2 x 2	4 x 4	8 x 8	16 x16	32 x 32	64 x 64	128x128	256x128
columns									
Mean firing rate	27	24	26	23	22	23	20	22	19
(Hz)									
Used cores1	1-8	1-32	1-	1-128	1-128	1-128	4-128	64-128	64-128
(min-max)			128						
MPI processes	1-8	1-32	1-	1–256	1-256	1-256	4-256	64-256	128
			128						
Execution time2	0.15	0.4	1.80	3.05	6.85	20.0	59	211	386
(execution sec /									
simulated sec)									
Normalized	2.73	5.36	2.41	4.22	6.0	4.22	3.61	2.94	3.07
execution	×10 ⁻⁸	×10 ⁻⁹	×10 ⁻⁸	×10 ⁻⁹					
time3:									
execution time /									
(firing rate ×									
total syn ×									
simulated									
second)									

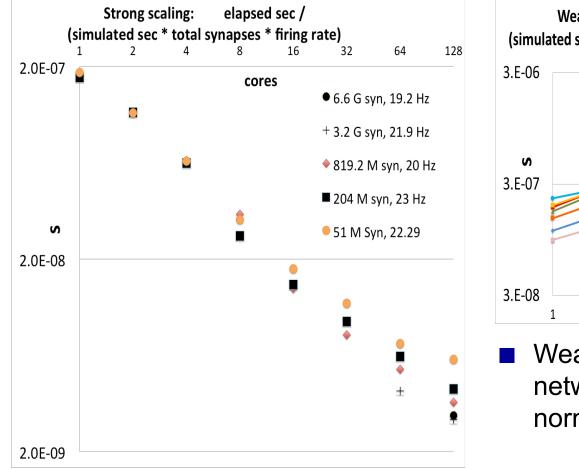
the "Strong scaling" section for a discussion about the unit of measure for the normalized execution speed.



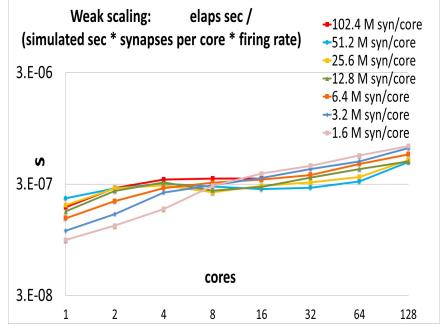
¹ Each cores @2.4 Ghz part of a quad-core Intel(R) Xeon(R) E5620.

² Using the "max number of cores reported in this table

DPSNN-STDP: MPI version - Strong and Weak Scaling



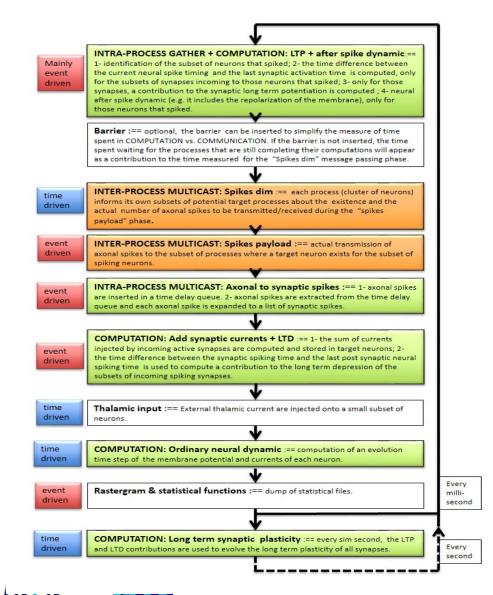
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Weak scaling for various local network sizes. Exec time normalized to synapse count.

Strong scaling. From 1 to 128 cores @ 2.4 GHz simulate various total network sizes (from 51 Mega syn to 6.6 Giga synapses). Exec time normalized to synapse count.

From Program Flow and Profiling ...



Function of the block	Relative	Note	
	execution		
	time		
Long term potentiation + after	(9.7 ±0.7)%	Gather1 +	
spike dynamic		computation	
Barrier (optional)2	(29.9±6.1)%	Workload	
		fluctuations	
Communication: inter-process	(0.77±0.10)%	Message	
multicast: Spikes dim		passing	
Communication: inter-process	(0.82±0.20)%	Message	
multicast: Spikes payload		passing	
Axonal to synaptic spikes: intra-	(16.8±2.3)%	Dereferencing3	
process multicast			
Add synaptic currents + long	$(19.2 \pm 2.7)\%$	Computation	
term depression			
Thalamic input4	0.01%	Simplified	
		model	
Ordinary neural dynamic	(11.8±1.4)%	Computation	
Rastergram & other statistical	(1.9±0.1)%	Computation	
functions			
Long term synaptic plasticity	(9.2±1.8)%	Computation	

1 The benchmarked software implementation is based on sparse accesses from the target neuron to the global list of incoming synapses. In a hardware implementation, based on several independent memory banks, if all synapses incoming to the same neuron were stored in contiguity, this task could be easily accelerated.

4 In this simulation the thalamic input is computed using a simple statistical model. Actually, this is one of the interface between the neural network and the "external" world, so its weight would greatly increase and add to that of other interfaces to be added.

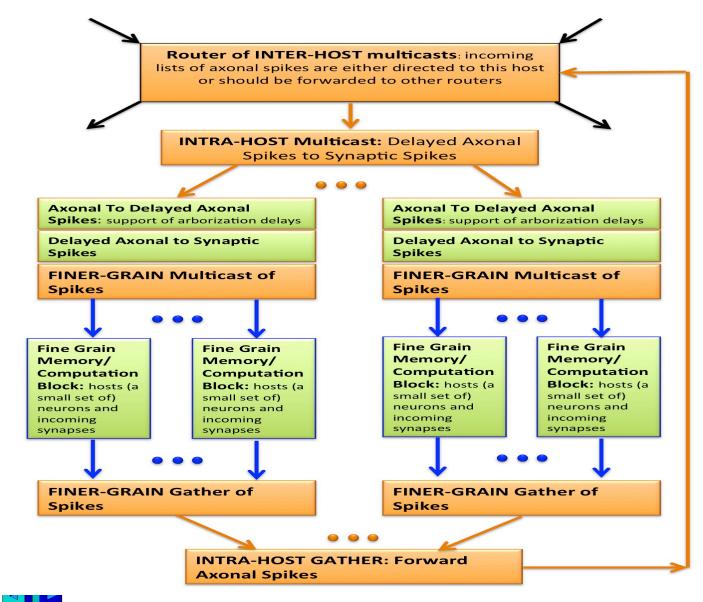




² If the barrier is not inserted, the time spent waiting for the processes that are still completing their computations appear as a contribution to the time measured for the "spikes dim" communication phase. We verified that the deviations from ideal strong scaling, can be entirely attributed to the cost of measured fluctuations in workload execution (represented by the Barrier block) and to (a very small) cost of communications.

³ The task to be performed is an "intra-process" multicast, from axons to specific lists of synapses. Instead, the benchmarked software implementation is based on two levels of dereferencing.

... toward Hardware Acceleration



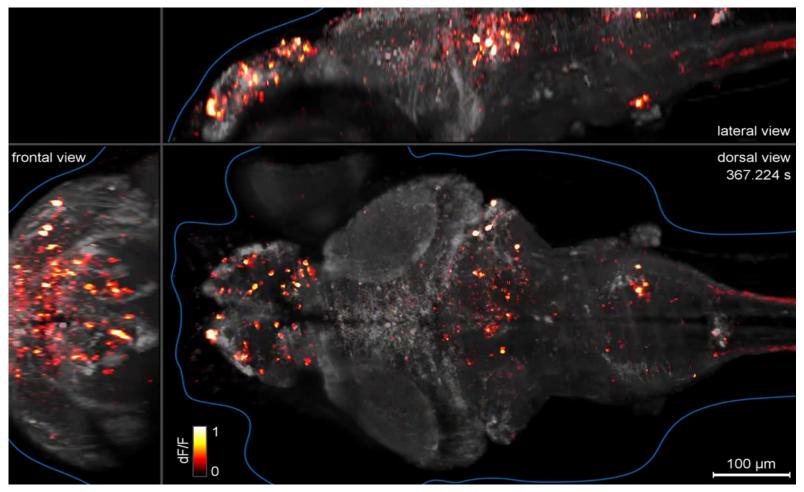
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CORTICONIC FET project

- Experimental techniques used to stimulate and measure cortical activities on animals and humans: opto-genetic, electrode arrays, trans-cranial magnetic stimulation, electro encephalographic arrays, drug perfusion...
- Large scale simulations
- DPSNN will be improved to simulate biological networks:
 - Cooperation between INFN and ISS
- Comparison with in-vivo/in-vitro experimental results
 - IDIBAPS (Barcellona), TUM



Spiking activity of individual neurons observed in real-time (e.g. in a Zebra Fish Larva)



Misha B Ahrens, Philipp J Keller, «Whole-brain functional imaging at cellular resolution using lightsheet microscopy», Nature Methods, 18 March 2013, DOI:10.1038/NMETH.2434 Howard Hughes Medical Institute, 3D recording of temporal spiking activity of ~100 000 neurons. Note: the effective time resolution is still only ~1 s.





Hardware-Software Co-design opportunity

- Huge potential for architectural improvements driven by this benchmark:
 - the brain performs with 50 W computations that would require more than 50 MW on present generation HPC architectures
- Strategic research area...



Conclusion

- present INFN activity on cortical/brain modeling
 - Coding of scalable Parallel/Distributed simulator
 - Comparison with experimental neuro-biological data and calibration of the INFN simulator
 - Interface with experimental systems / inclusion of the simulator into robotic platforms
 - Co-design of simulation code and execution platform
- A strategic Large Scale Computing research theme...

