Exploiting network structure in NEST: Efficient communication in brain-scale simulations

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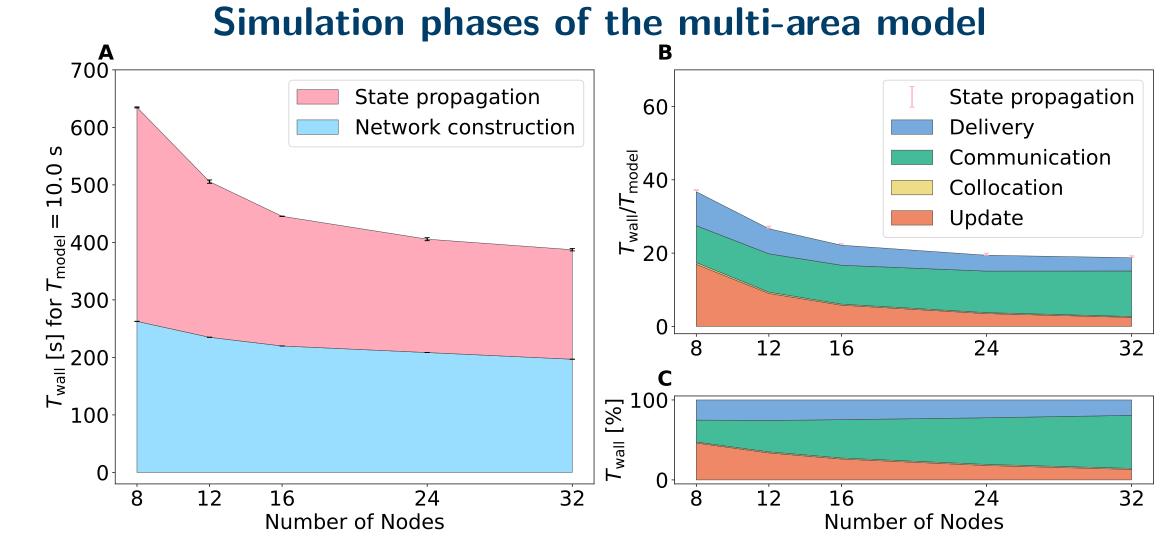






Motivation

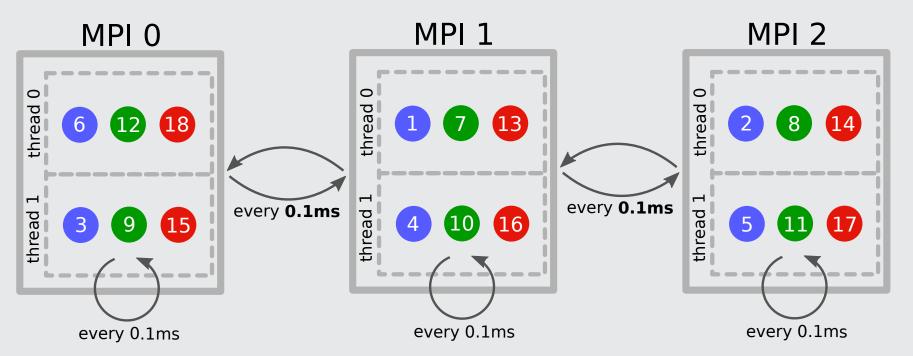
- continuous improvement of CPU-based simulation techniques create challenging benchmarking targets for neuromorphic platforms
- neuronal simulations on conventional hardware still maintain higher flexibility at potentially lower cost compared to novel dedicated hardware [1]
- spike communication is the bottleneck in simulations of brain-scale networks [2]
- e.g. the multi-area model of macaque visual cortex [3]
 - 32 interconnected areas modelled as microcircuits [4]
 - realistic connectivity
 - single neuron resolution
- ⇒ structure-aware neuron distribution scheme combined with optimized spike-communication framework to speed up neuronal simulations



Strong-scaling benchmark of macaque multi-area model performed with NEST v3.6 on Jülich Supercomputer JURECA.

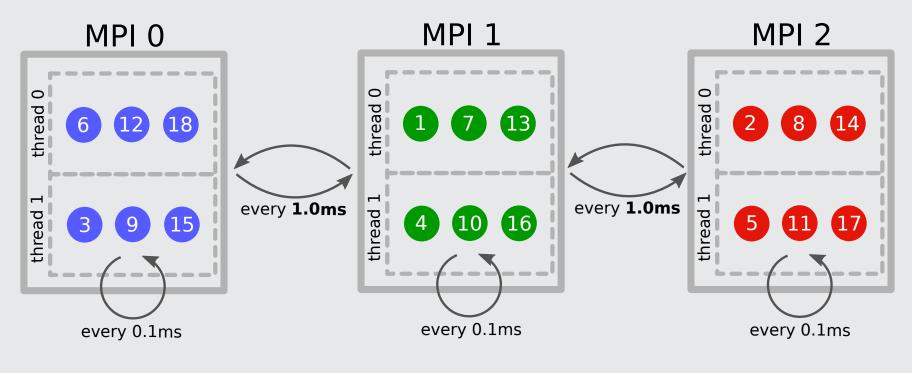
Algorithm

Conventional neuron distribution scheme



- uniform occupation of compute nodes: "round-robin"
- neurons of the same area are spread out on the hardware
- \blacksquare communication between compute nodes every smallest delay of e.g. $0.1\,\mathrm{ms}$

Structure-aware neuron distribution scheme

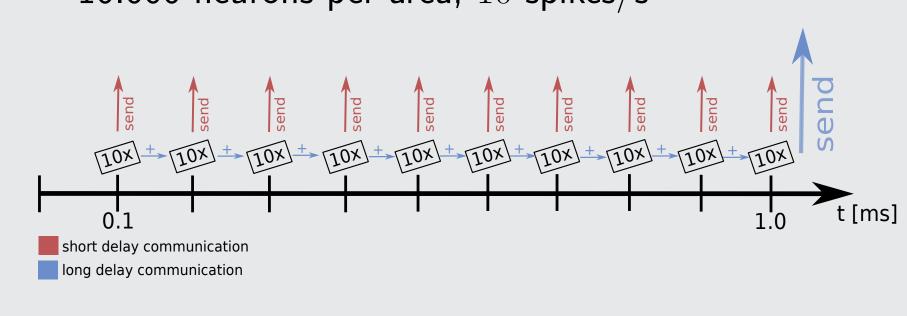


- one/few compute nodes per area
- two communication pathways
- within an area: short delays (e.g. $0.1 \,\mathrm{ms}$)
- between areas: long delays (e.g. $1.0 \,\mathrm{ms}$)
- \Rightarrow faster communication within areas
- ⇒ fewer communication between areas

>1 ms

Example: structure-aware approach

■ 10.000 neurons per area; 10 spikes/s



Results

Setup

- neuronal simulator tool NEST [5,6]
- benchmarking model
- similar to macaque multi-area model in connectivity and work load
- easily scalable while retaining constant activity levels
- $-\approx 130.000$ neurons per area
- $\, \approx \, 3000$ inter- and intra-area connections per neuron, respectively
- average spike rate of 2.5 spike/s
- Jülich Supercomputer JURECA
 - -2 areas per compute node
- communication phase
- synchronization between all compute nodes (only long-range communication)
- spike data exchange (both short-range and longrange communication)

Weak-scaling experiment State propagation Delivery Communication Synchronization Collocation Update 10 0 8 16 32 64 8 16 32 64

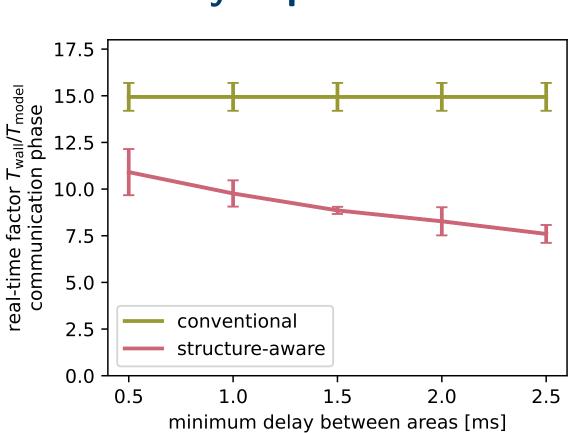
A Conventional round-robin neuron distribution with single communication pathway. **B** Structure-aware neuron distribution with separate communication pathways for short- and long-range connections

significant speed up of spike communication

Number of Nodes

- benefit lies in reduced time spent on compute node synchronization
- promising scaling behavior for large number of areas and compute nodes

Delay dependence



- delay distr. within an area: $\mathcal{N}(1.25, 0.625)$
- delay distr. between areas: $\mathcal{N}(5.00, 2.50)$
- lower cutoff of inter-area delay distribution defines inter-node communication frequency in structure-aware approach
- ⇒ benefit of implementation increases with decreasing amount of inter-area communication

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Outlook

Number of Nodes

- benchmarking of networks with inhomogeneous activity or size
- benchmarking state of the art models (e.g. multi-area model of macaque visual cortex)

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