

Exploiting network topology in brain-scale multi-area model simulations

Melissa Lober^{1,2*}, Markus Diesmann^{1,3}, Susanne Kunkel⁴

¹ Institute of Neuroscience and Medicine (INM-6), Institute for Advanced Simulation (IAS-6) and JARA Brain Institute I (INM-10), Jülich Research Centre, Jülich, Germany

² RWTH Aachen University, Aachen, Germany

³ Department of Physics, Faculty 1, & Department of Psychiatry, Psychotherapy, and Psychosomatics, Medical School, RWTH Aachen University, Aachen, Germany

⁴ Neuromorphic Software Ecosystems (PGI-15), Jülich Research Centre, Jülich, Germany

* m.lober@fz-juelich.de

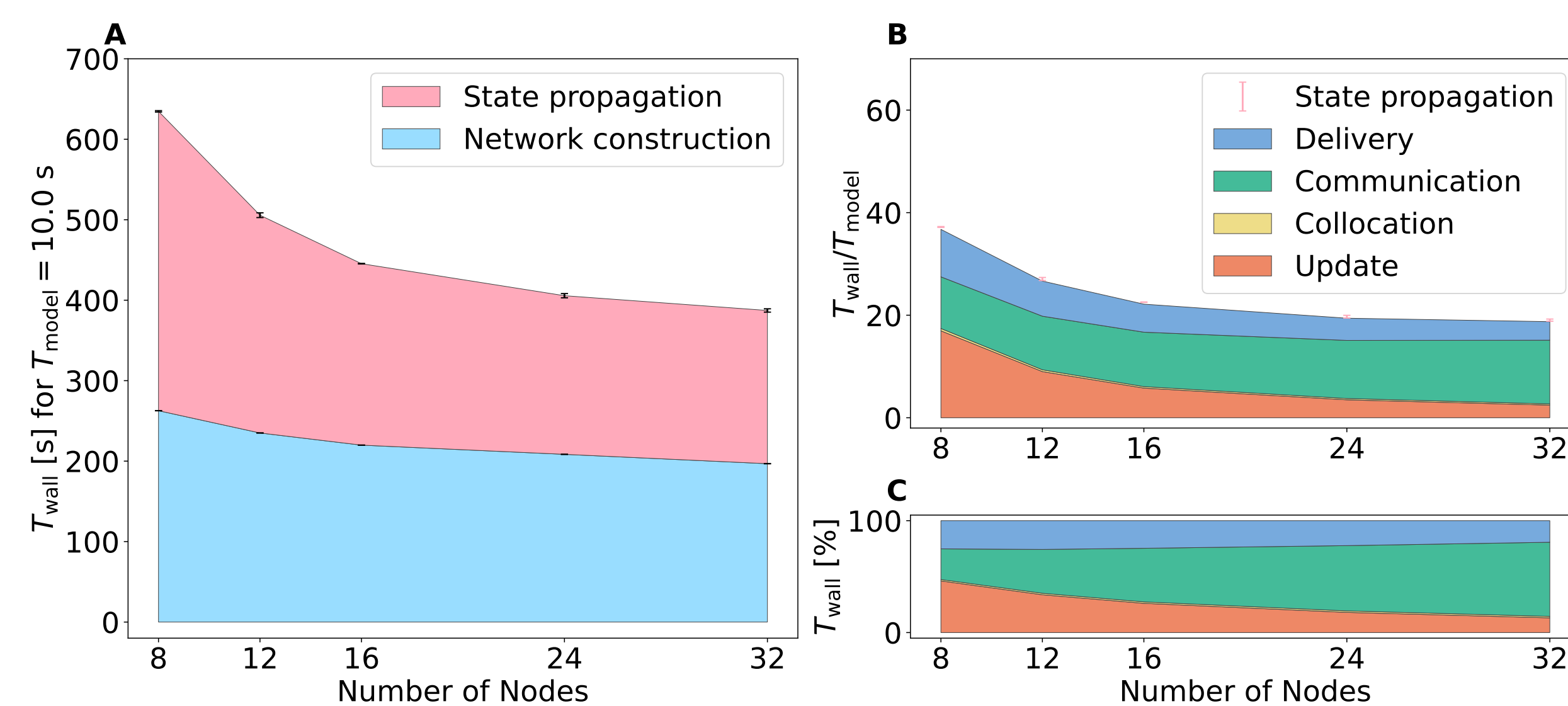


Motivation

- 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

Simulation phases of the multi-area model



Outlook

- benchmarking of networks with inhomogeneous activity or size
- benchmarking state of the art models (e.g. model of macaque visual cortex [3])

References

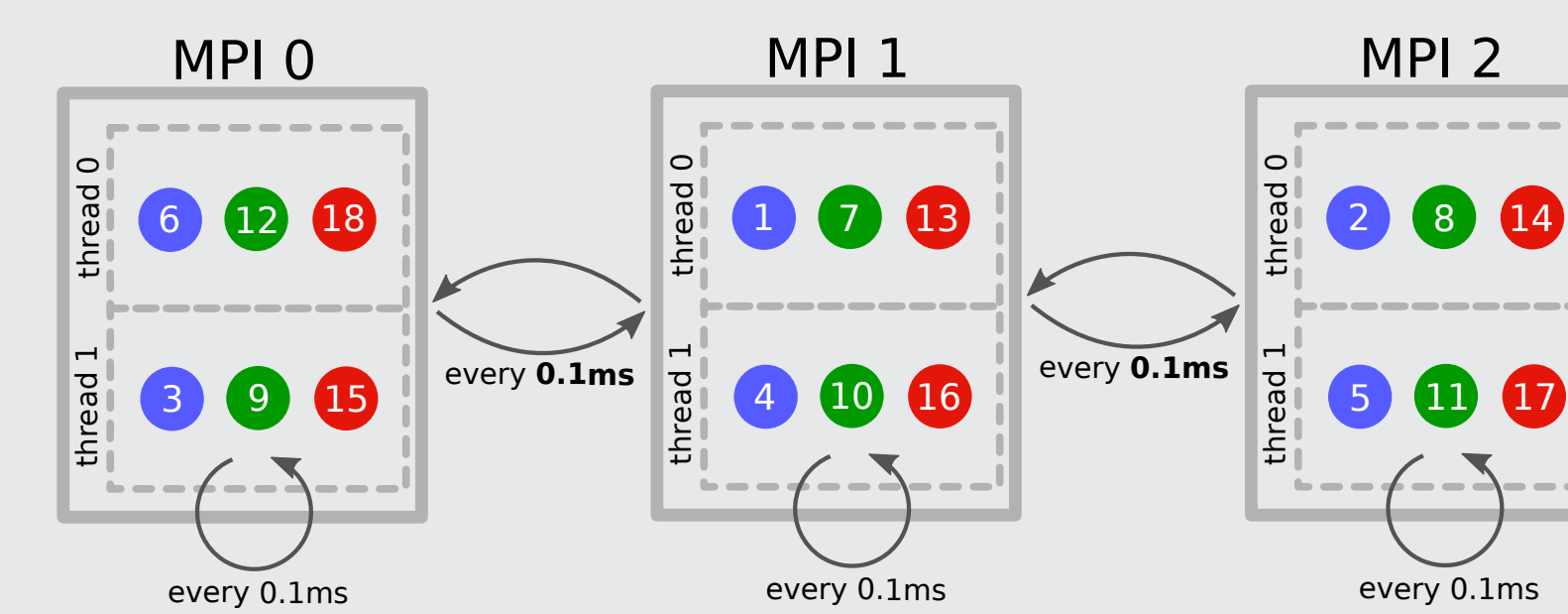
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Acknowledgments

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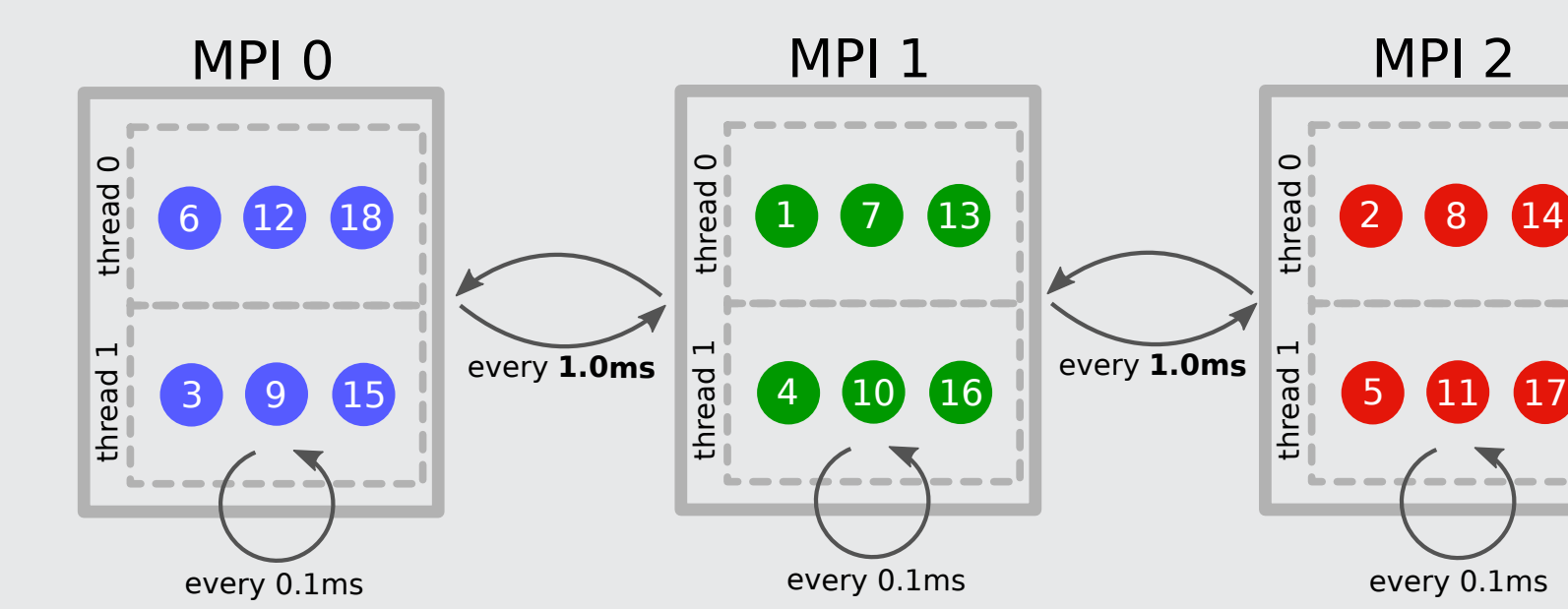
Algorithm

Conventional neuron distribution scheme

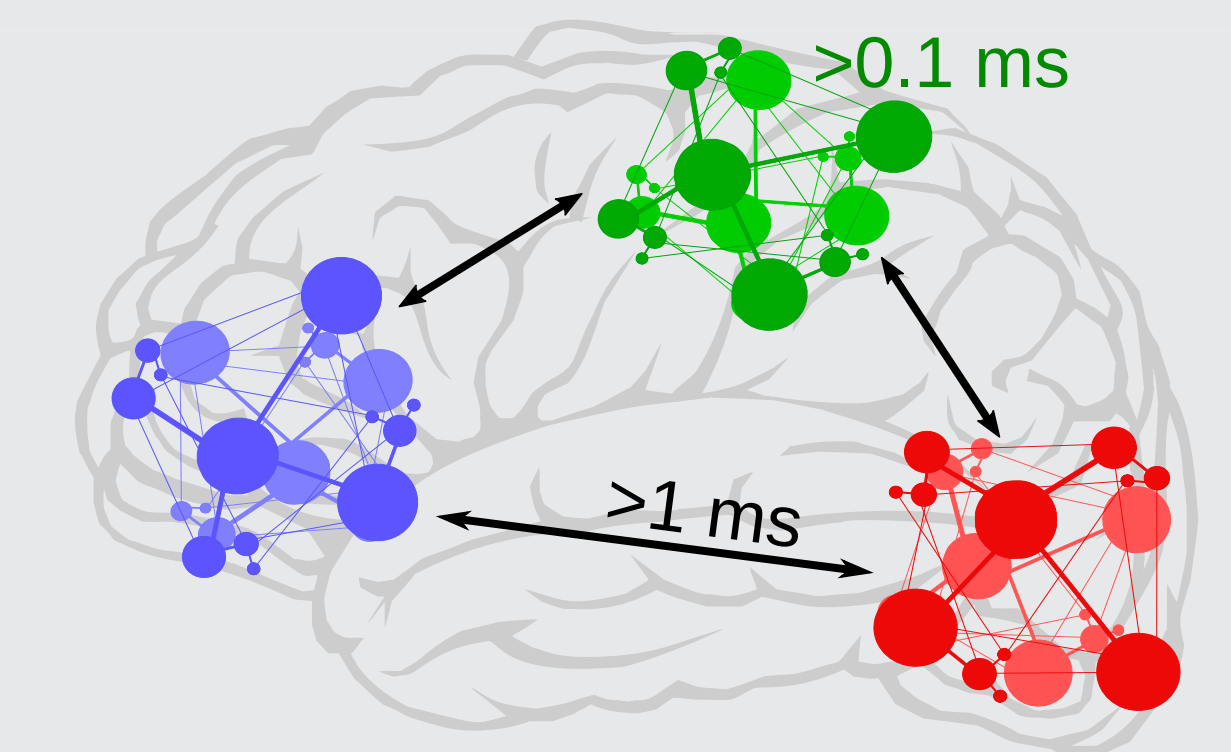


- uniform occupation of compute nodes: “round-robin”
- neurons of the same area are spread out on the hardware
- communication between compute nodes every smallest delay of e.g. 0.1 ms

Structure-aware neuron distribution scheme

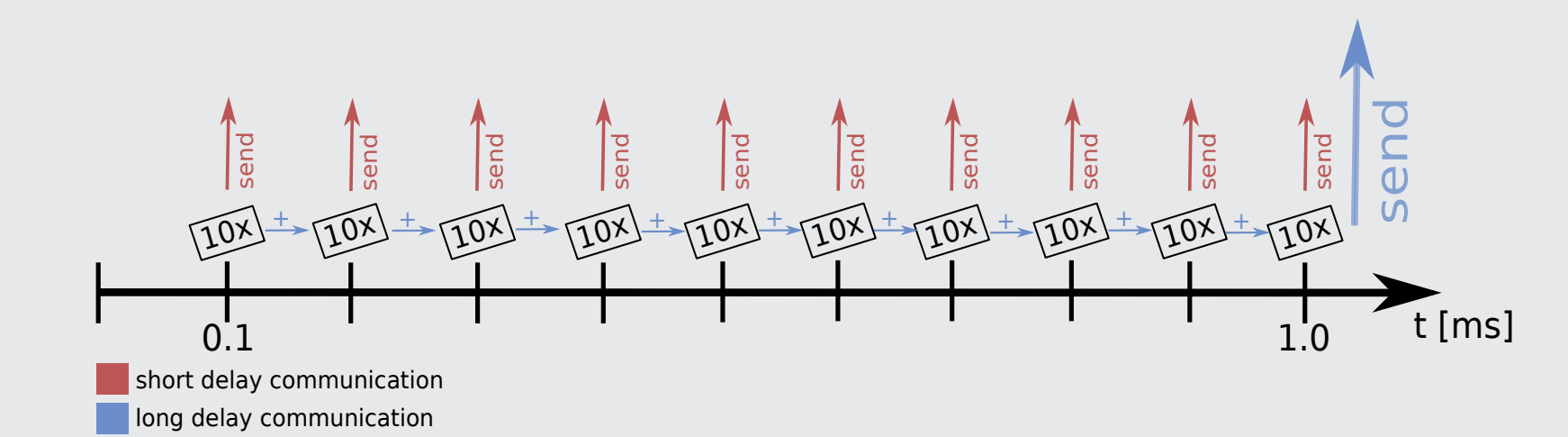


- one/few compute nodes per area
 - two communication pathways
 - within an area: short delays (e.g. 0.1 ms)
 - between areas: long delays (e.g. 1.0 ms)
- ⇒ faster communication within areas
- ⇒ fewer communication between areas



Example: structure-aware approach

- 10.000 neurons per area; 10 spikes/s

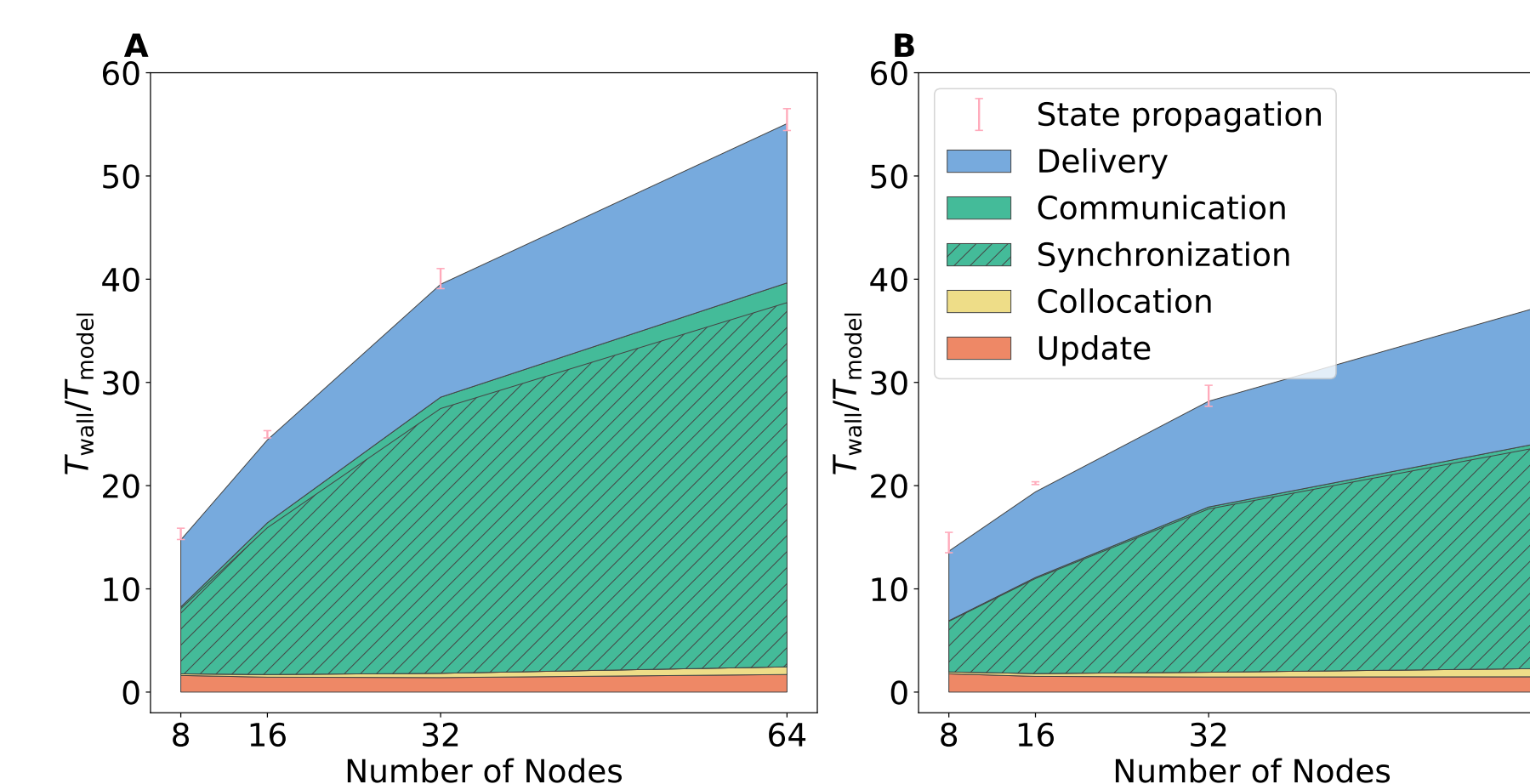


Results

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
 - ≈ 130.000 neurons per area
 - ≈ 3000 inter- and intra-area connections per neuron, respectively
 - average spike rate of 2.5 spike/s
- Jülich Supercomputer JURECA
 - 2 MPI processes per compute node, 64 threads each
 - 1 area per MPI process
- communication phase
 - synchronization between all compute nodes (only long-range communication)
 - spike data exchange (both short-range and long-range communication)

Weak-scaling experiment

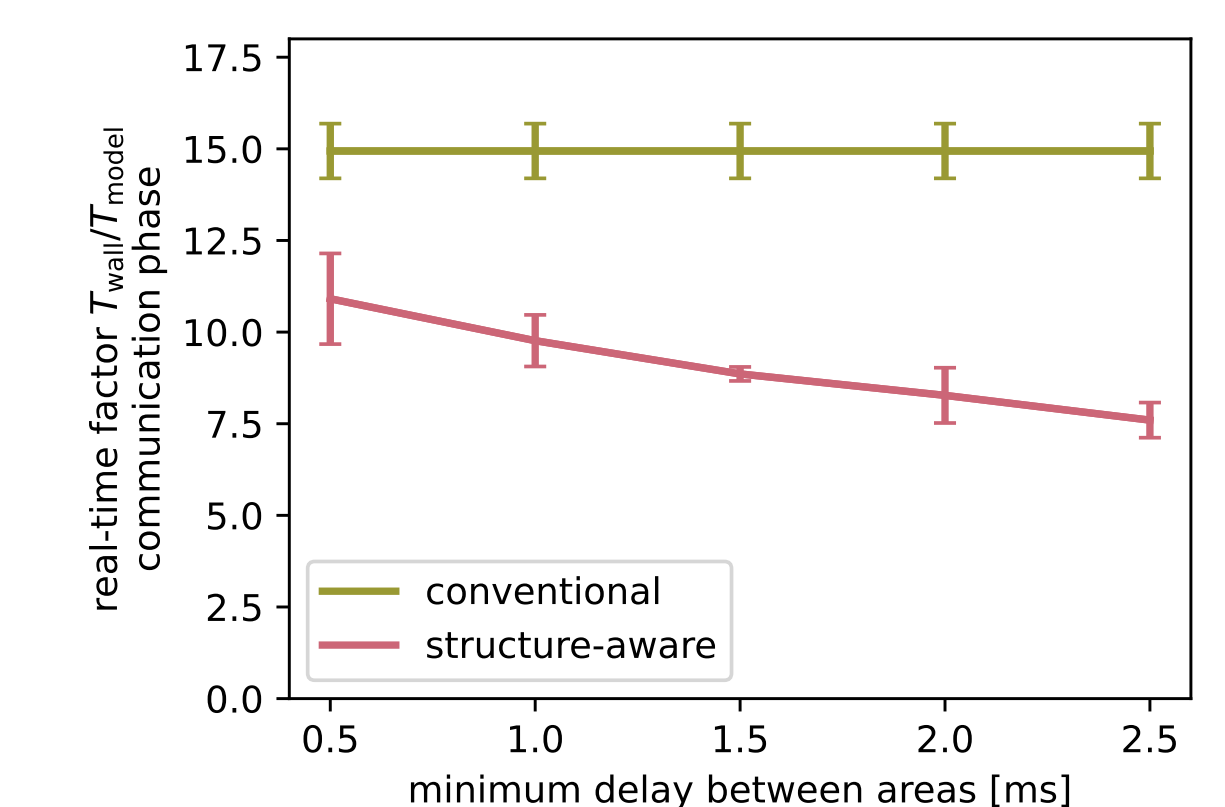


A Conventional round-robin neuron distribution with single communication pathway.

B Structure-aware neuron distribution with separate communication pathways for short- and long-range connections

- significant speed up of spike communication
- speed of other simulation phases is maintained
- 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 drawn from normal distribution $\mathcal{N}(1.25, 0.625)$
 - delay distr. between areas drawn from normal distribution $\mathcal{N}(5.00, 2.50)$
 - 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