Relaxed Scheduling for Scalable Belief Propagation

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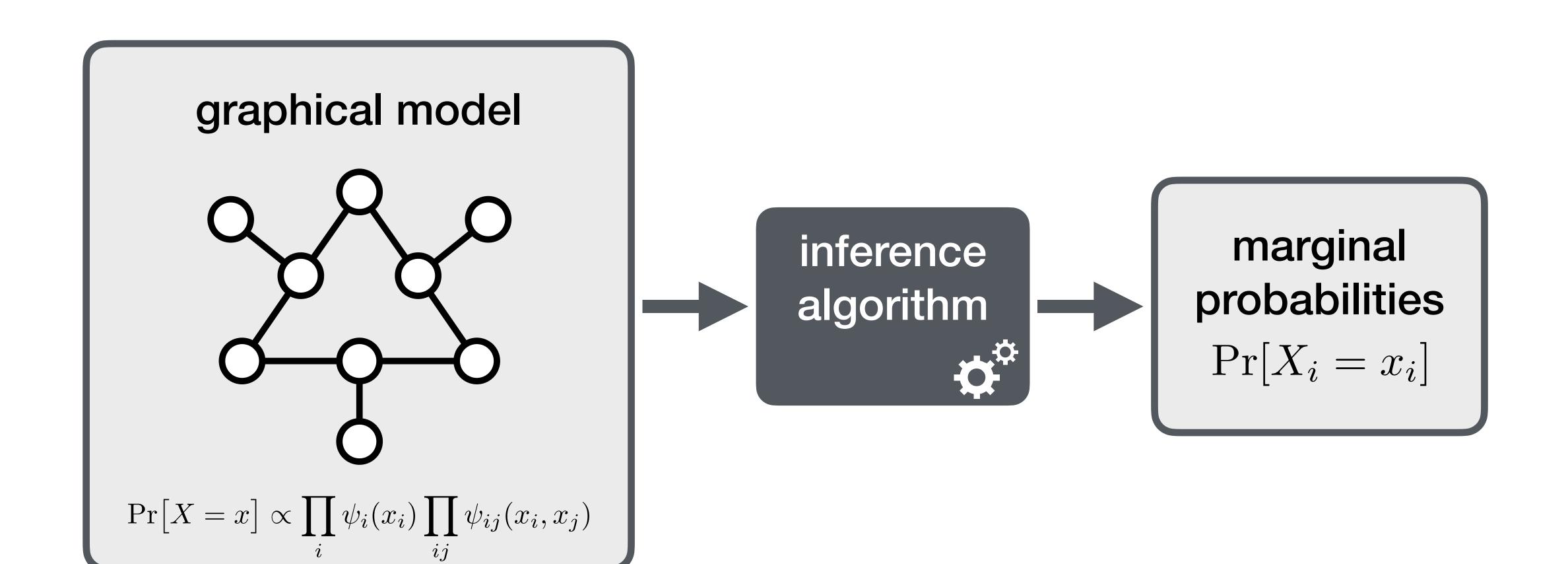
IST Austria

Joint work with

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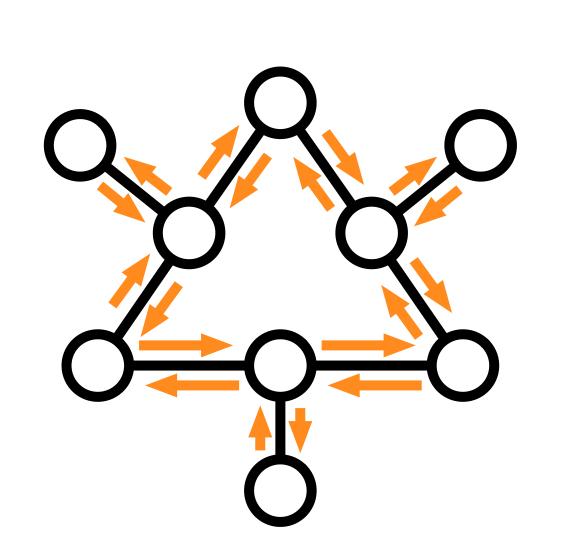
Inference on probabilistic graphical models



Parallelising belief propagation

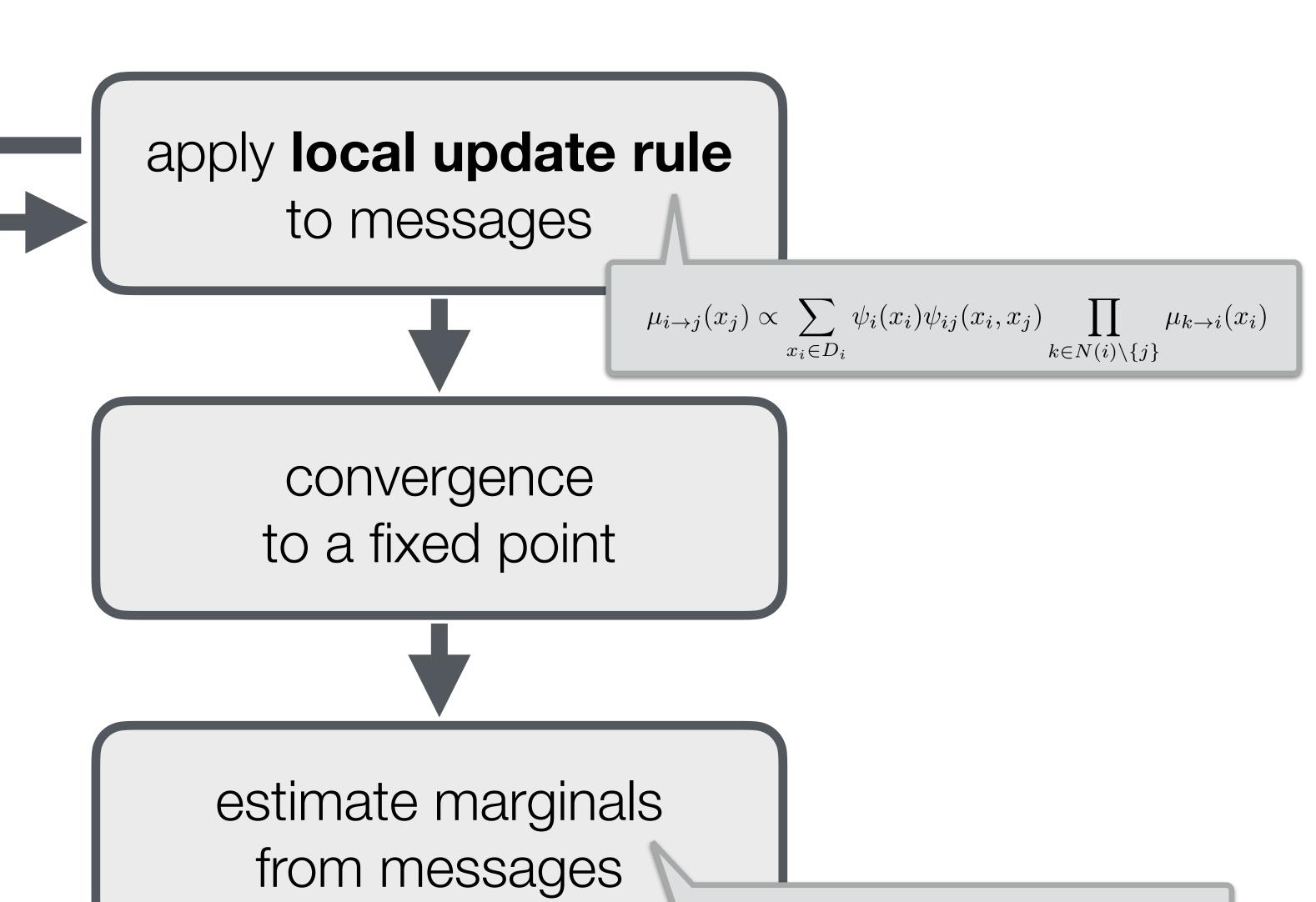
(in shared memory parallel setting)

Belief propagation



messages associated with edges of the graph

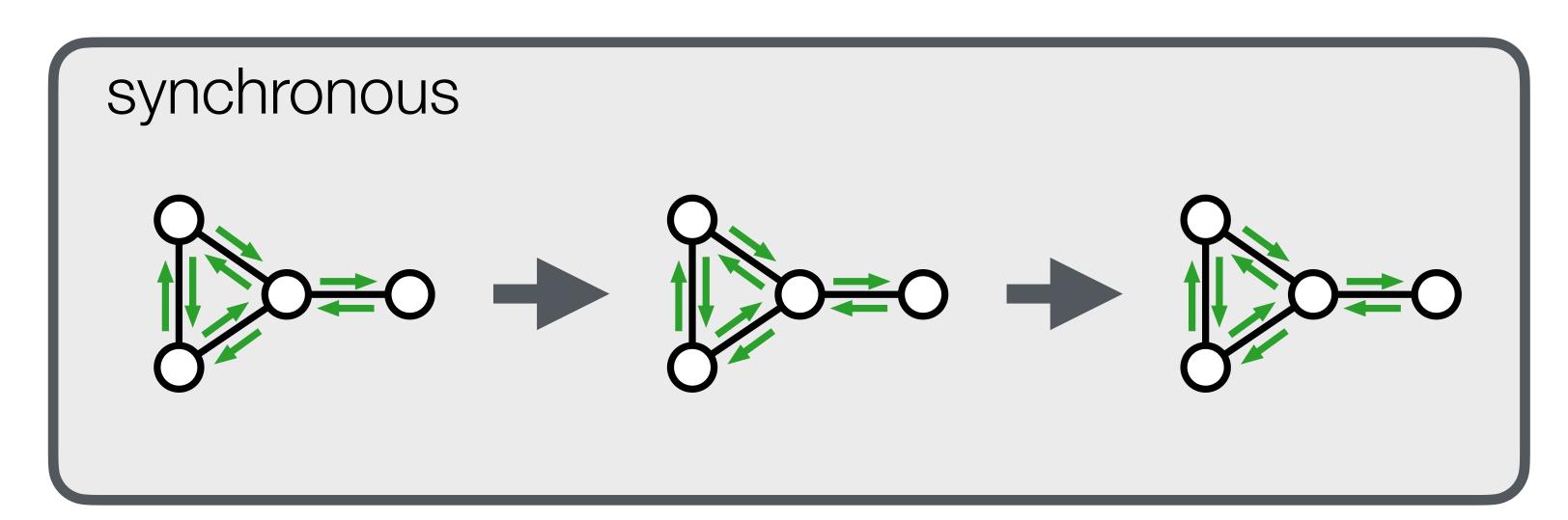
$$(\mu_{i\to j}\in\mathbb{R}^d)$$



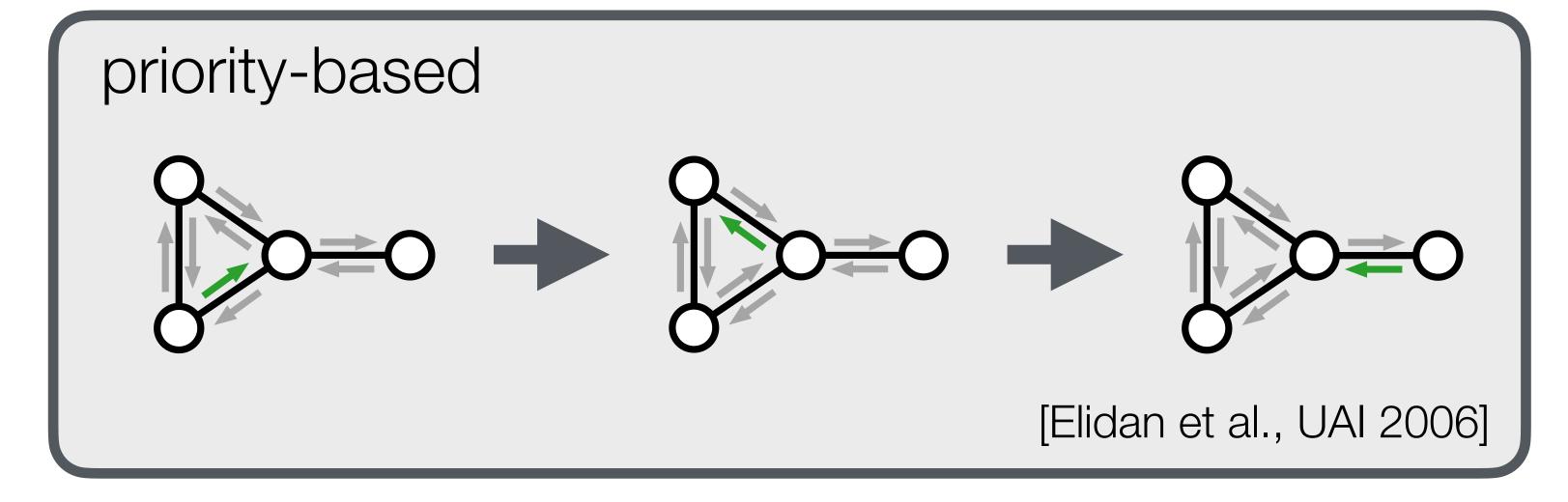
 $\Pr[X_i = x_i] \propto \psi_i(x_i) \prod \mu_{j \to i}(x_i)$

 $j \in N(i)$

Belief propagation message updates can be **scheduled** in any order

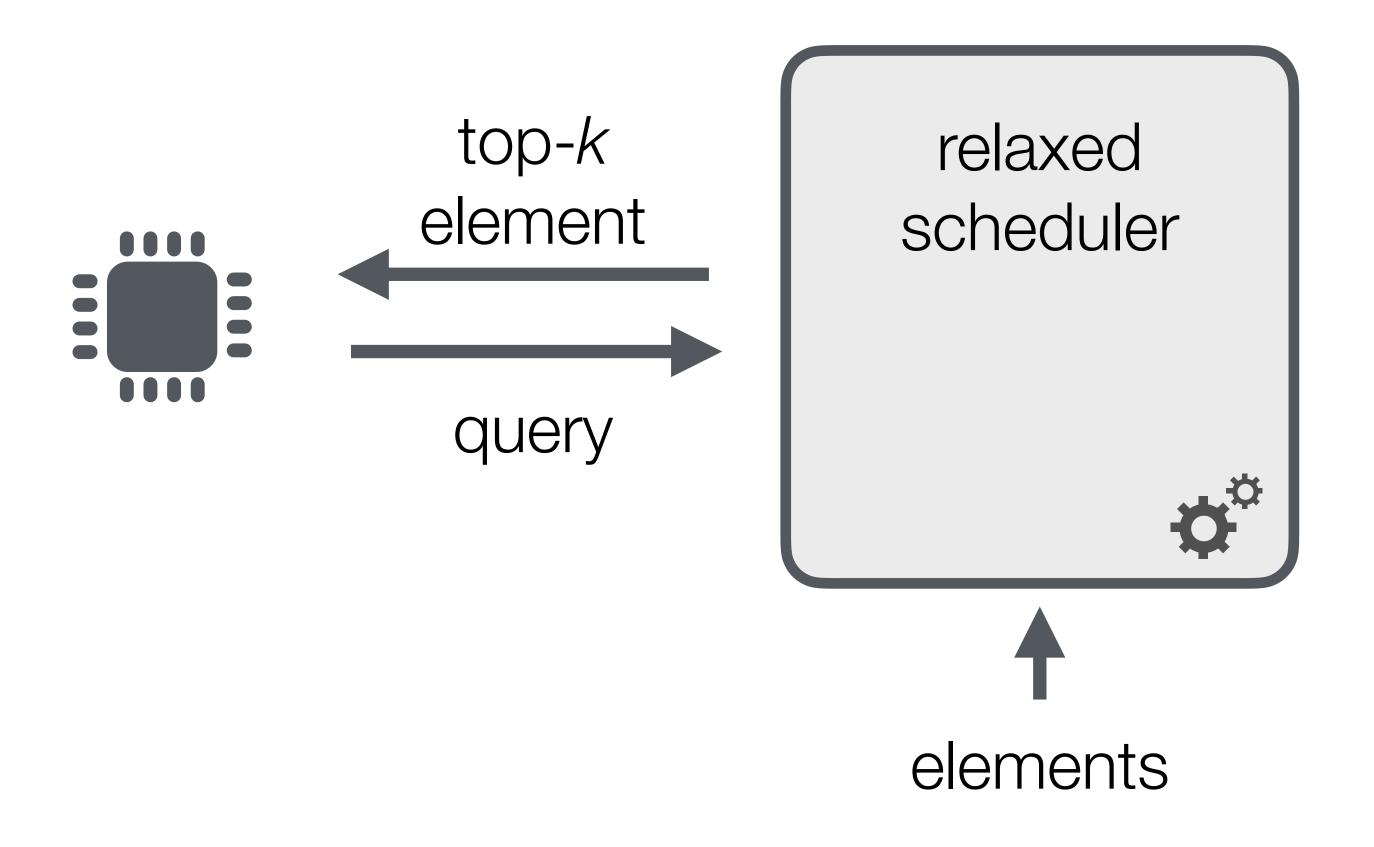


lots of updates easy to parallelise



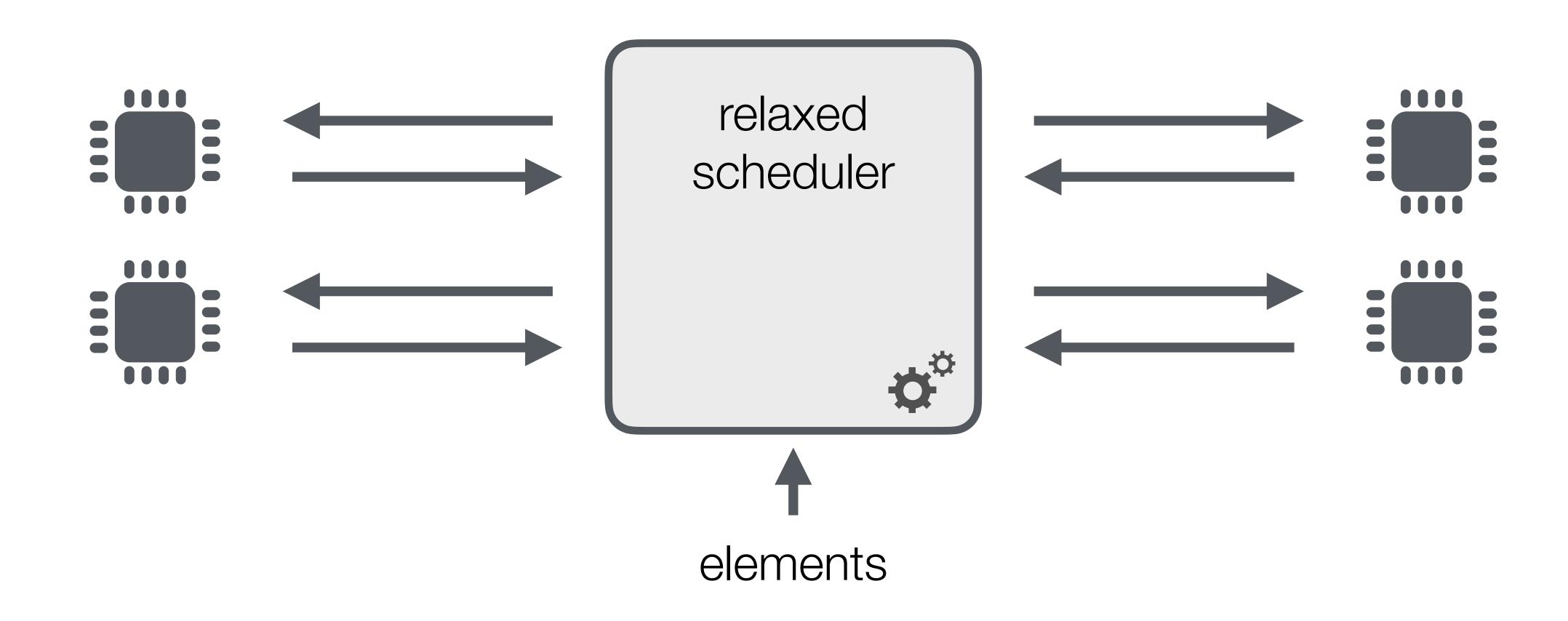
faster convergence fewer updates sequential, hard to parallelise

Priority-based belief propagation can parallelised efficiently using **relaxed schedulers**

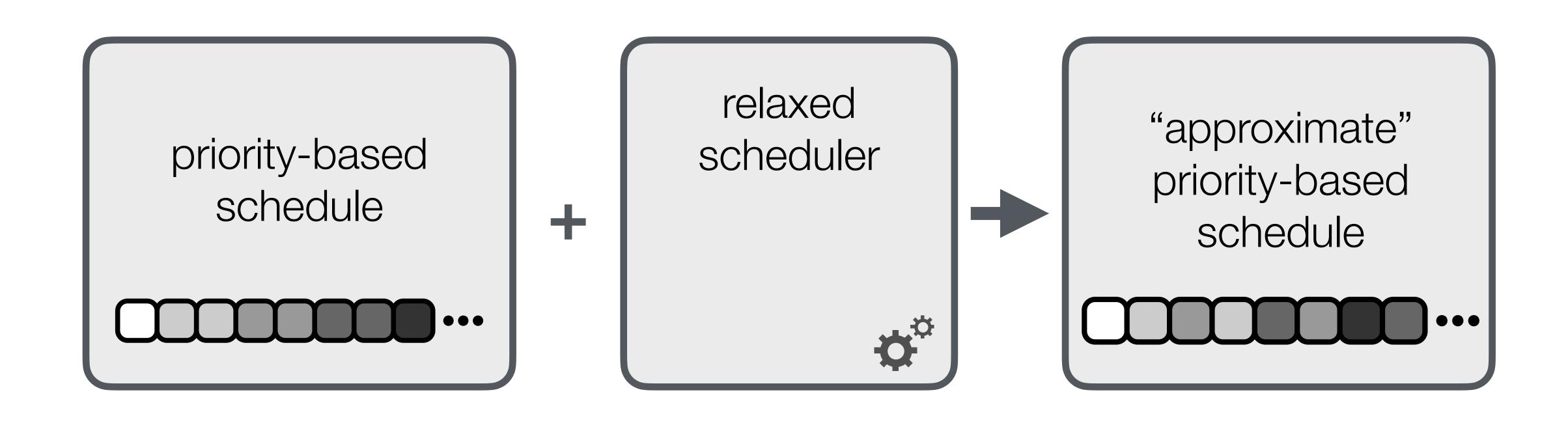


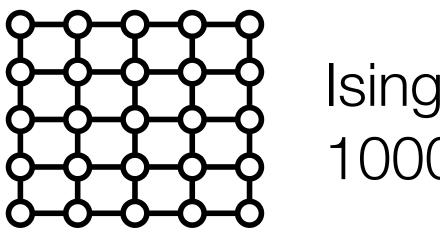
parallel data structure approximates a priority queue

Priority-based belief propagation can parallelised efficiently using **relaxed schedulers**



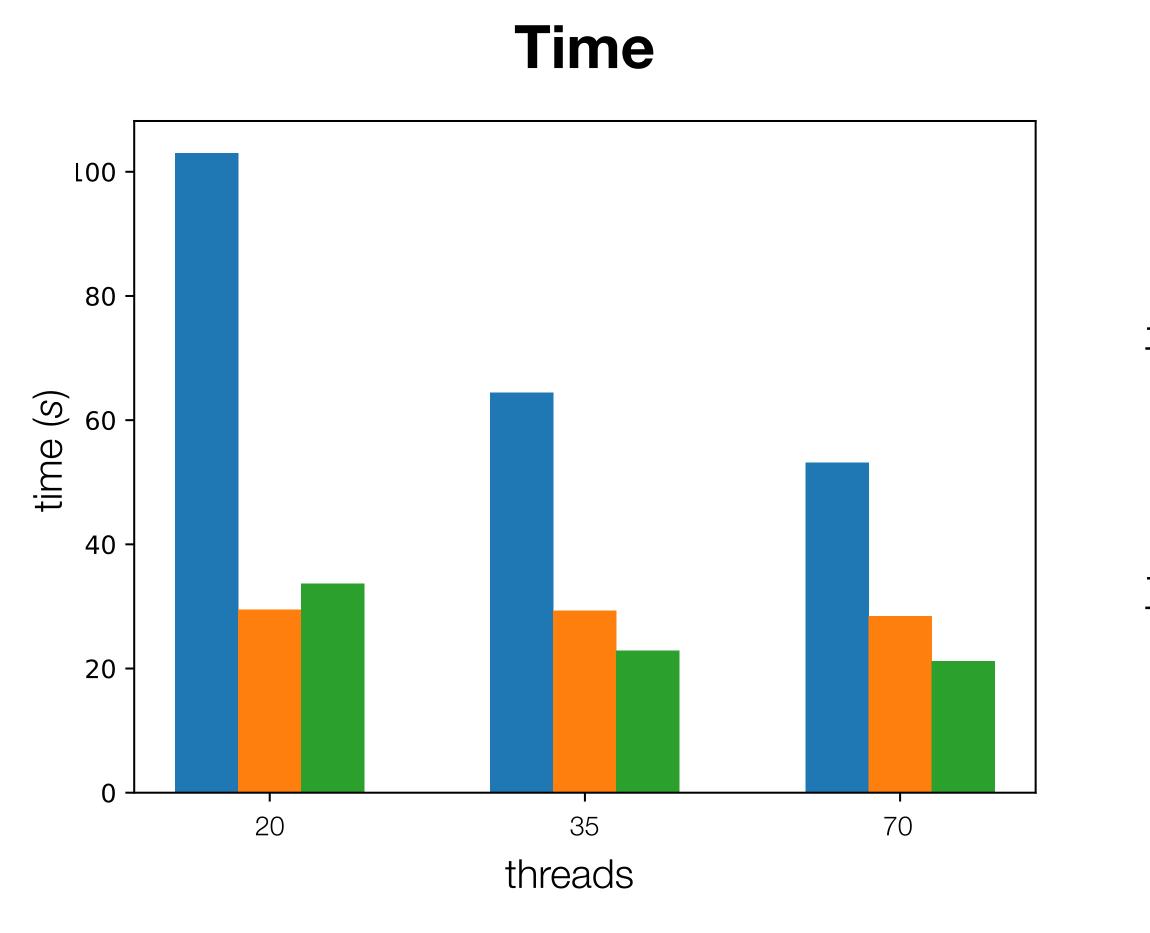
Priority-based belief propagation can parallelised efficiently using **relaxed schedulers**



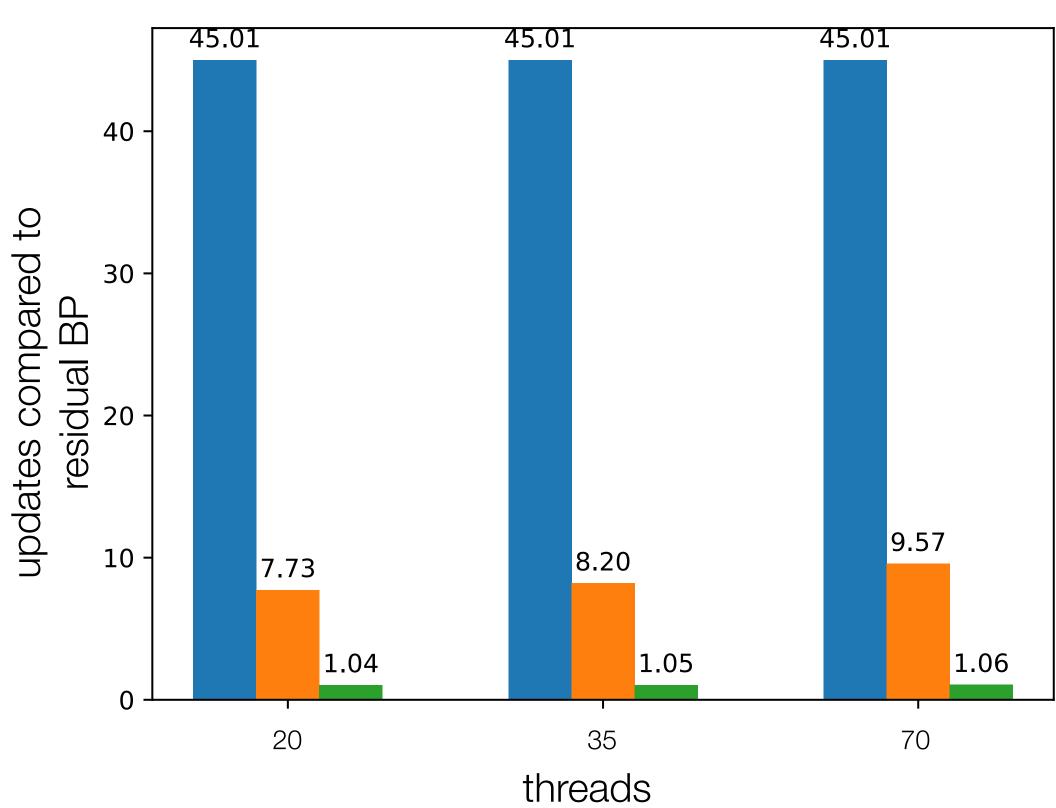


Ising model 1000×1000 grid

- synchronous
- residual splash [Gonzales et al, UAI 2009]
- relaxed residual (this work)



Number of updates

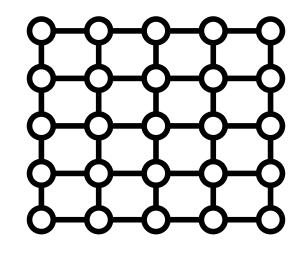


Simple parallel belief propagation implementation with state-of-the-art scaling

2

Relaxed schedulers are a powerful tool for parallelising iterative machine learning algorithms

arXiv:2002.11505



Ising model 1000×1000 grid

- synchronous
- residual splash [Gonzales et al, UAI 2009]
- relaxed residual (this work)



