

Gossip Algorithms for Majority Consensus

Emanuele Natale
www.enatale.name

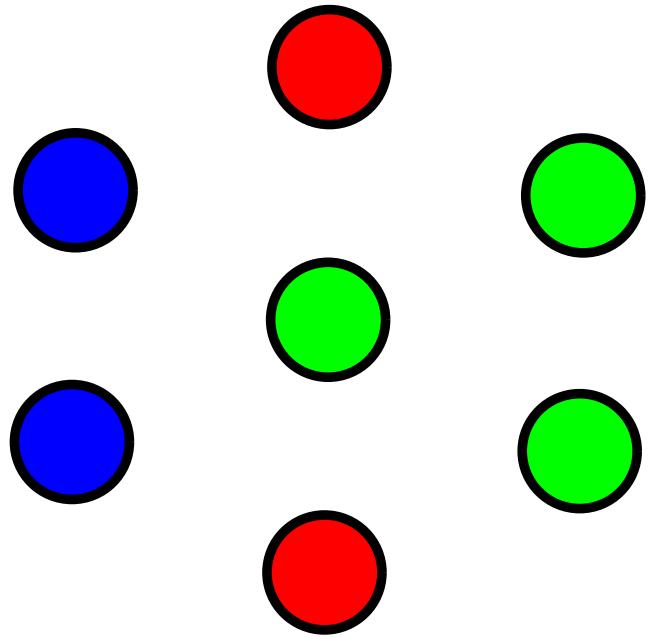
Supervisors: R. Silvestri, A. Clementi (Tor Vergata)

Research group: L. Becchetti, A. Clementi, F. Pasquale, R. Silvestri, (L. Trevisan) & me

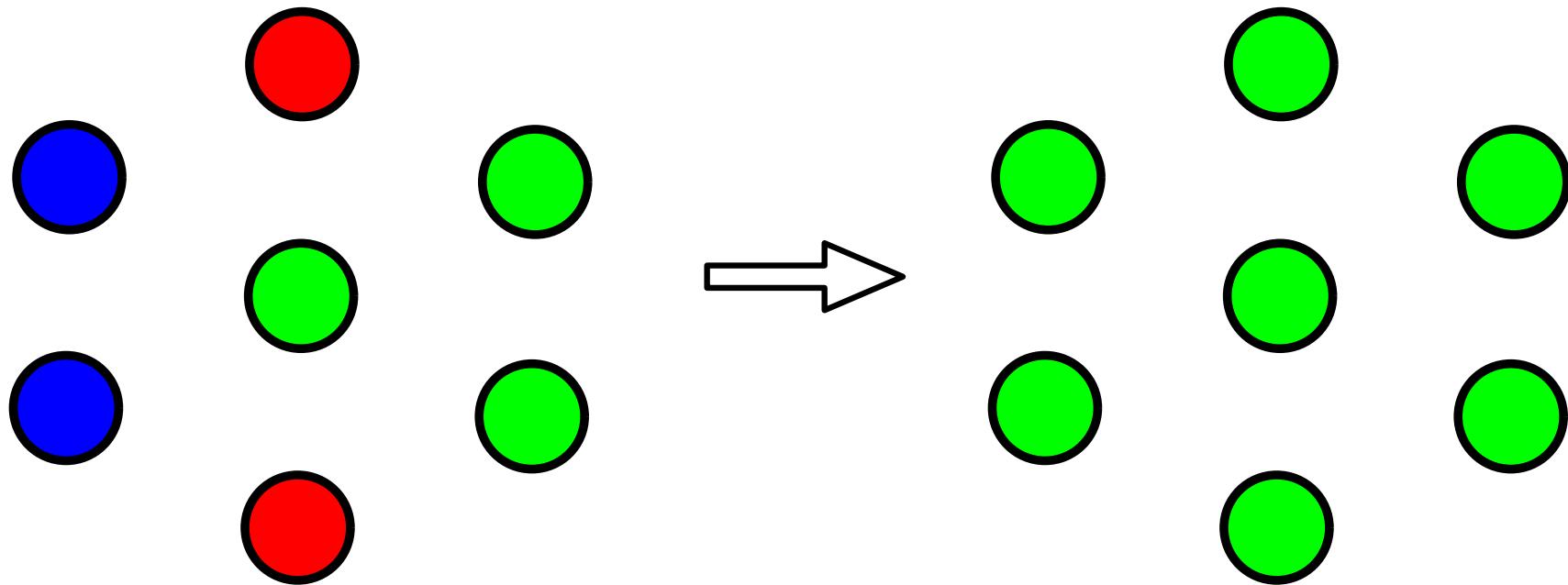


October 12, 2015

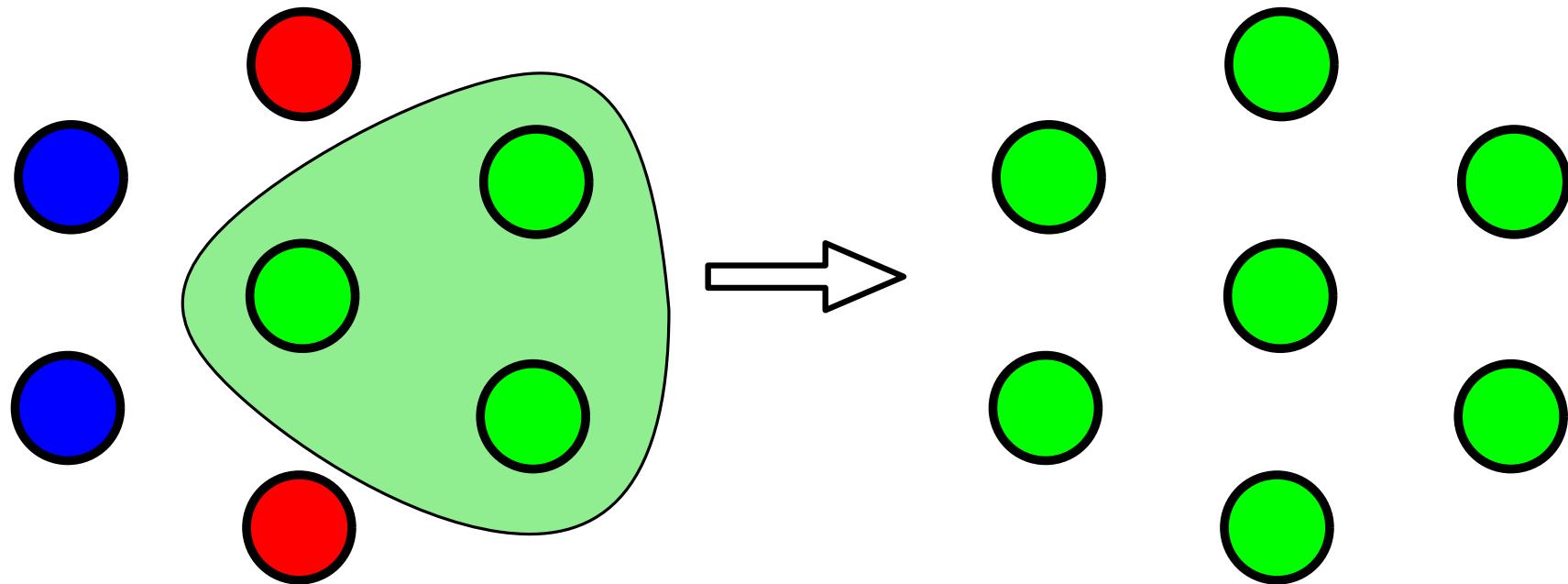
The Majority Consensus Problem



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$$3 > \max\{2, 2\}$$

Gossip Algorithms

Scenario: sensor networks, peer-to-peer networks,
mobile networks, vehicles networks...

⇒ Distributed, unstructured, dynamical, unreliable,
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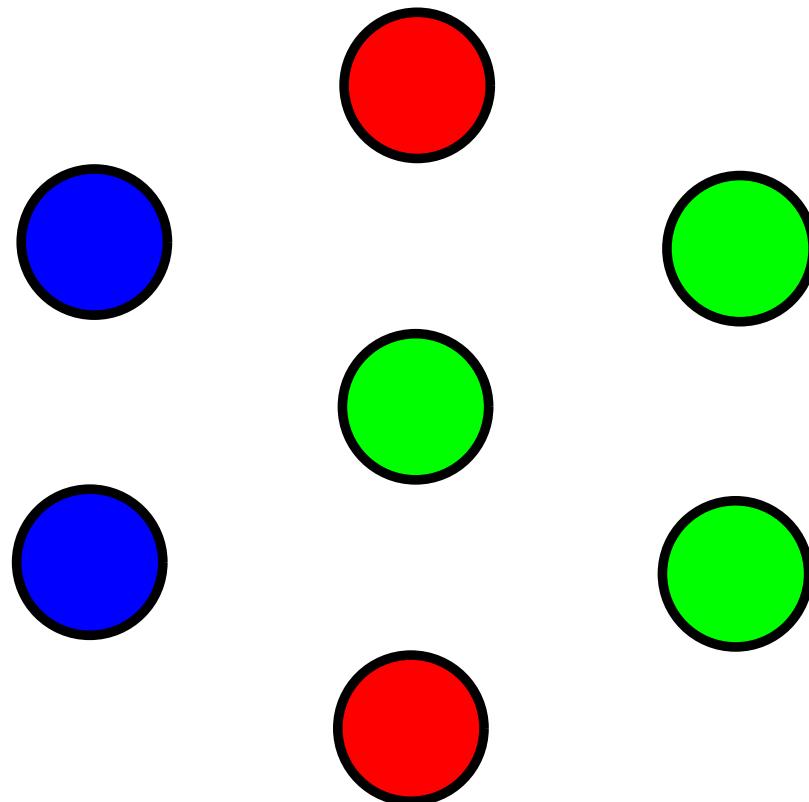
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Theoretical interest:

what can we do with minimal assumptions?

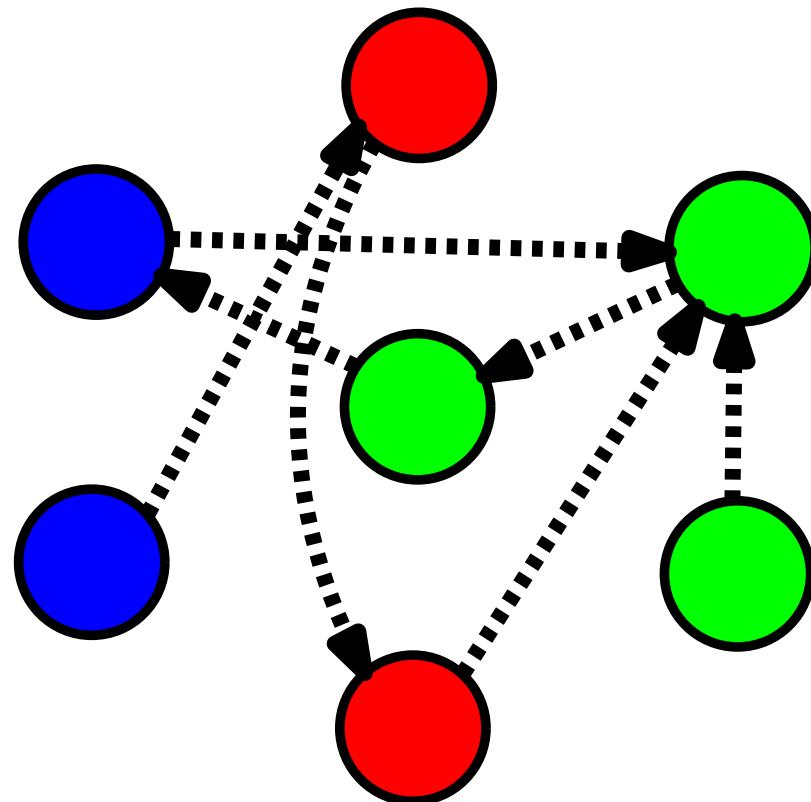
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At each round, each node can communicate with a finite number of nodes, chosen uniformly at random



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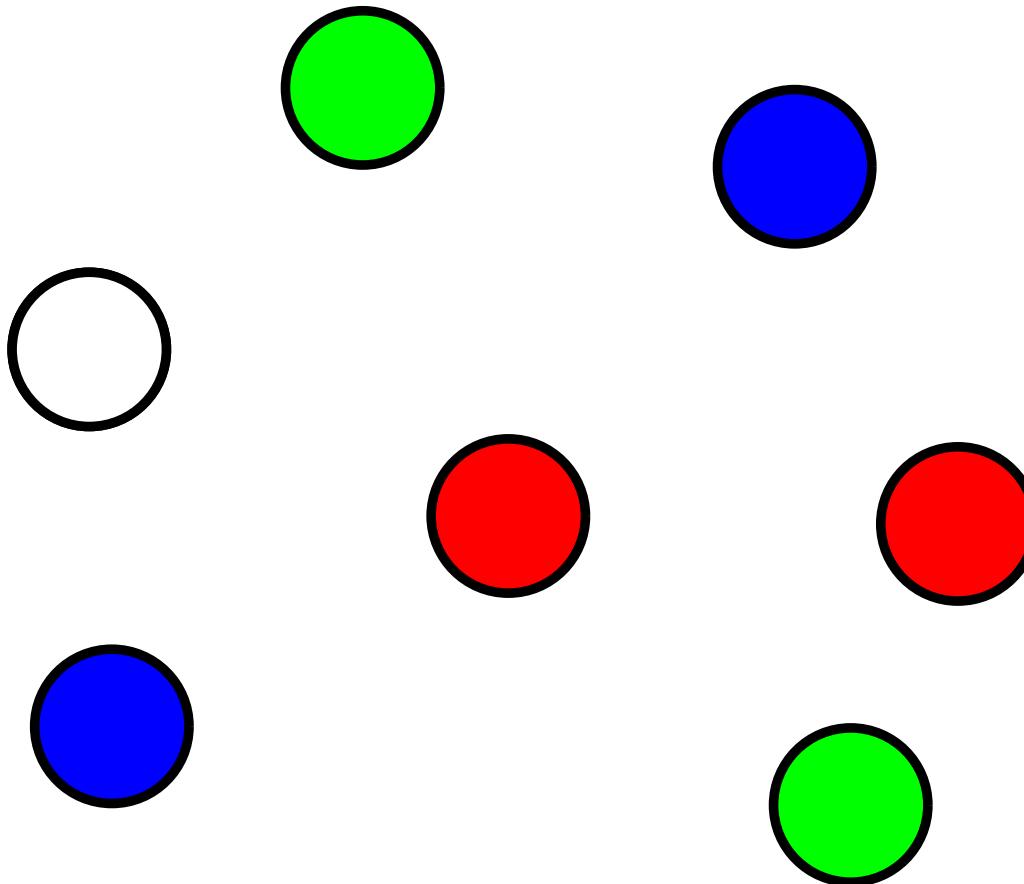
(Some) Related Works

	Mem. & mess. size	# of colors	Time efficiency	Comm. Model
Kempe et al. FOCS '03	$O(k \log n)$	any	$O(\log n)$	<i>GOSZIP</i>
Angluin et al. DISC '07 Perron et al. INFOCOM '09	$\Theta(1)$	2	$O(\log n)$	Sequential
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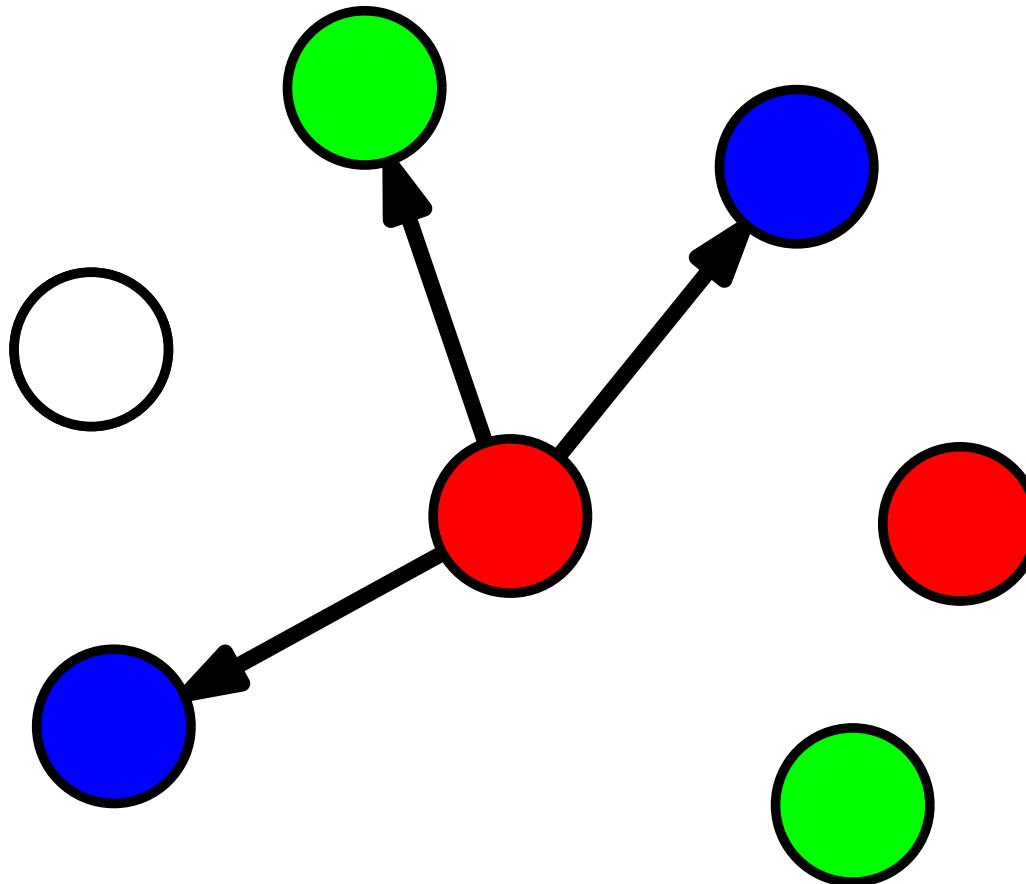
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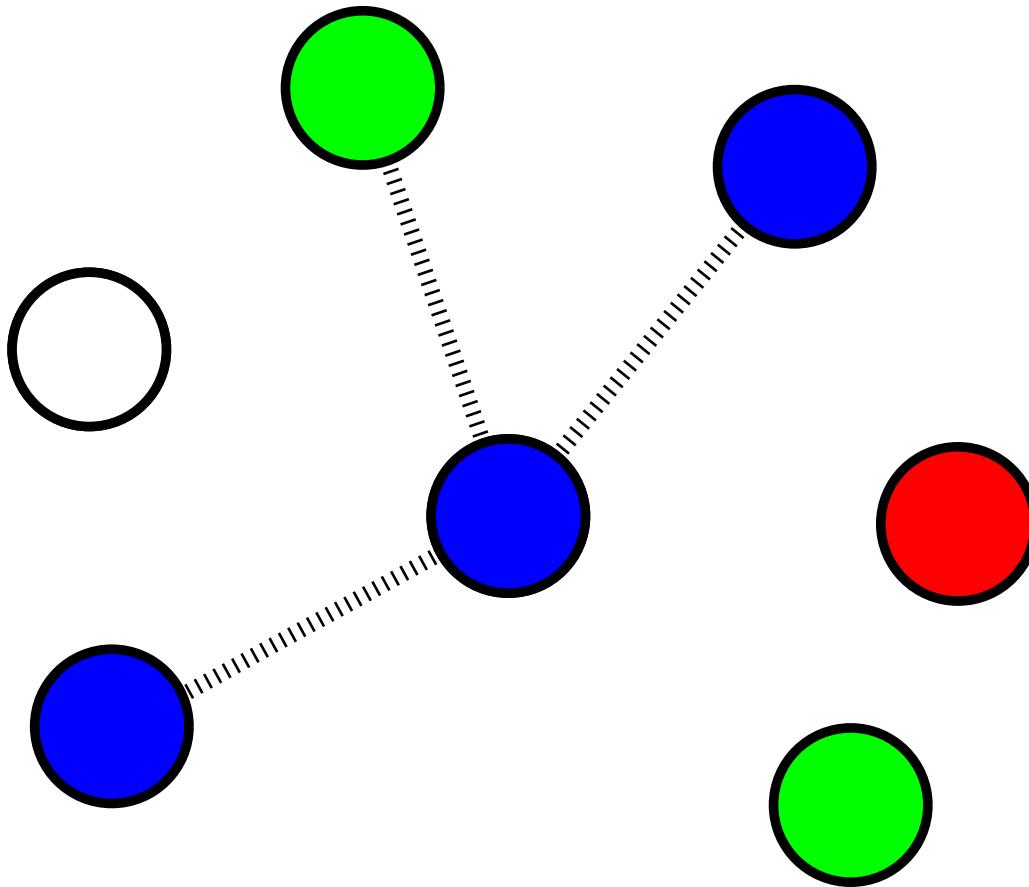
The 3-Majority Protocol



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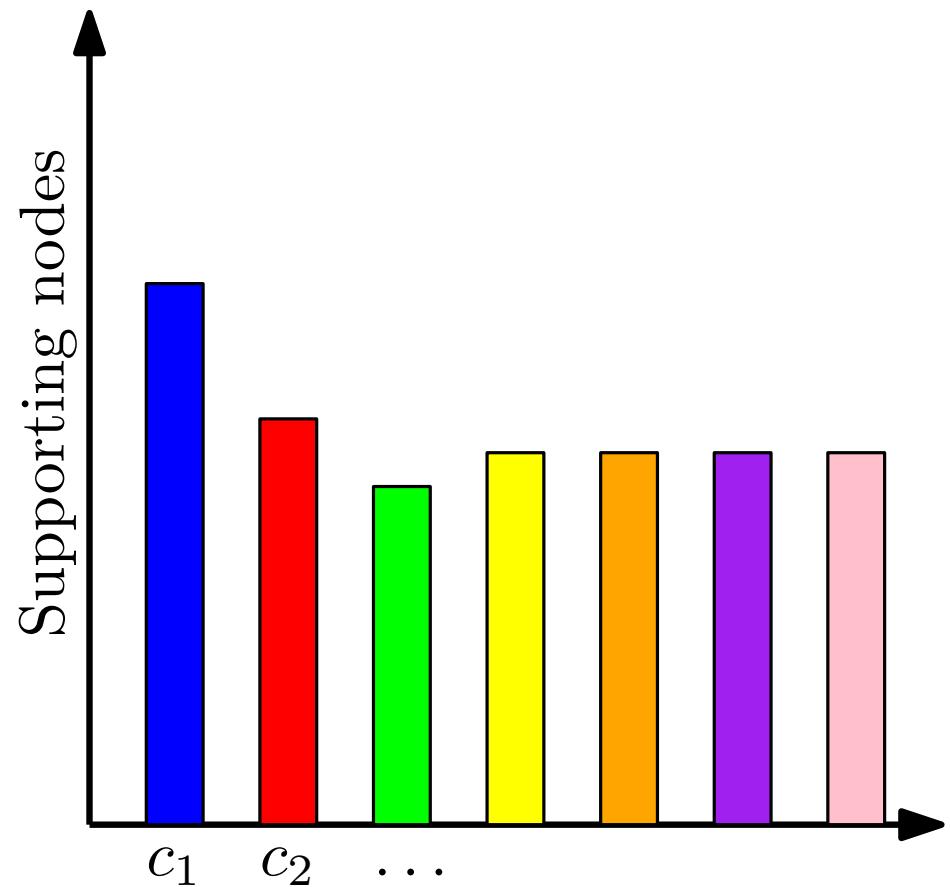
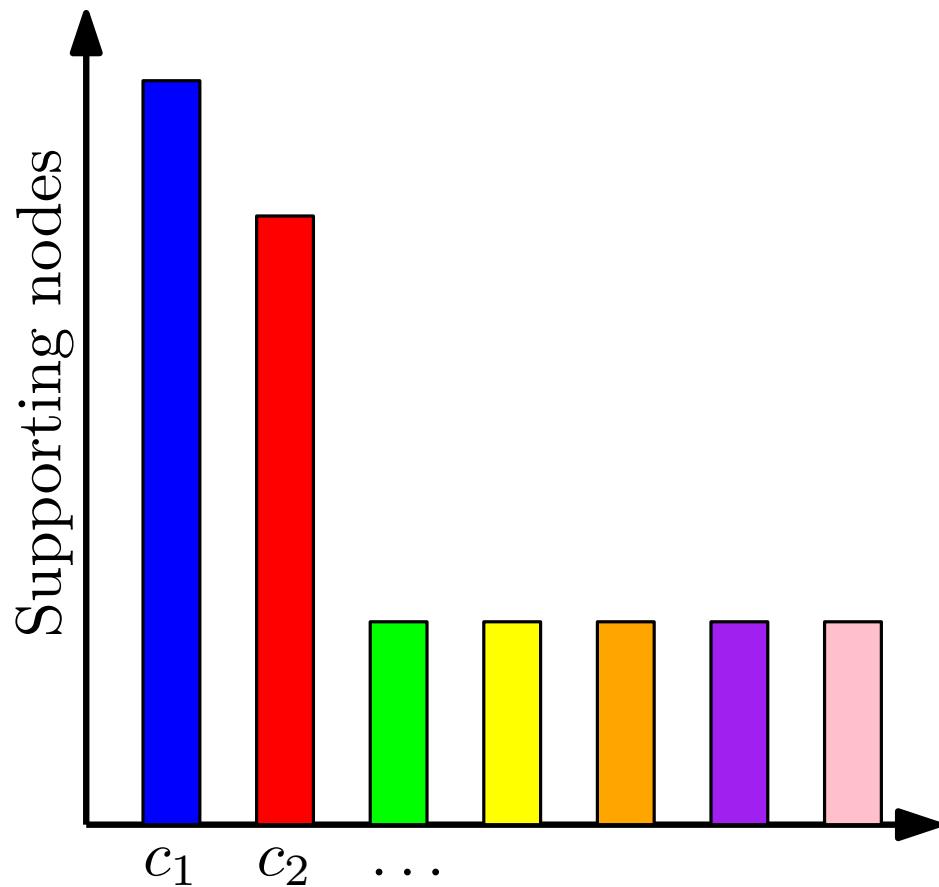
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Key Parameter of 3-Majority

$c_i^{(t)} := |\{i\text{-colored nodes}\}|$

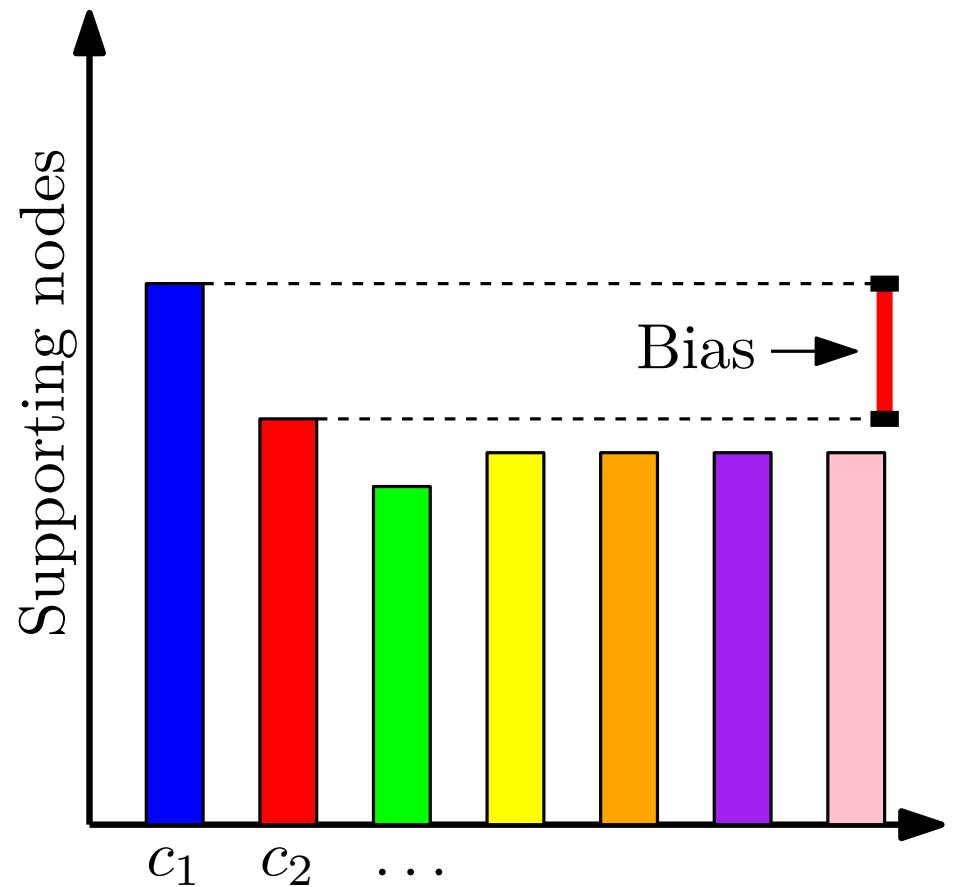
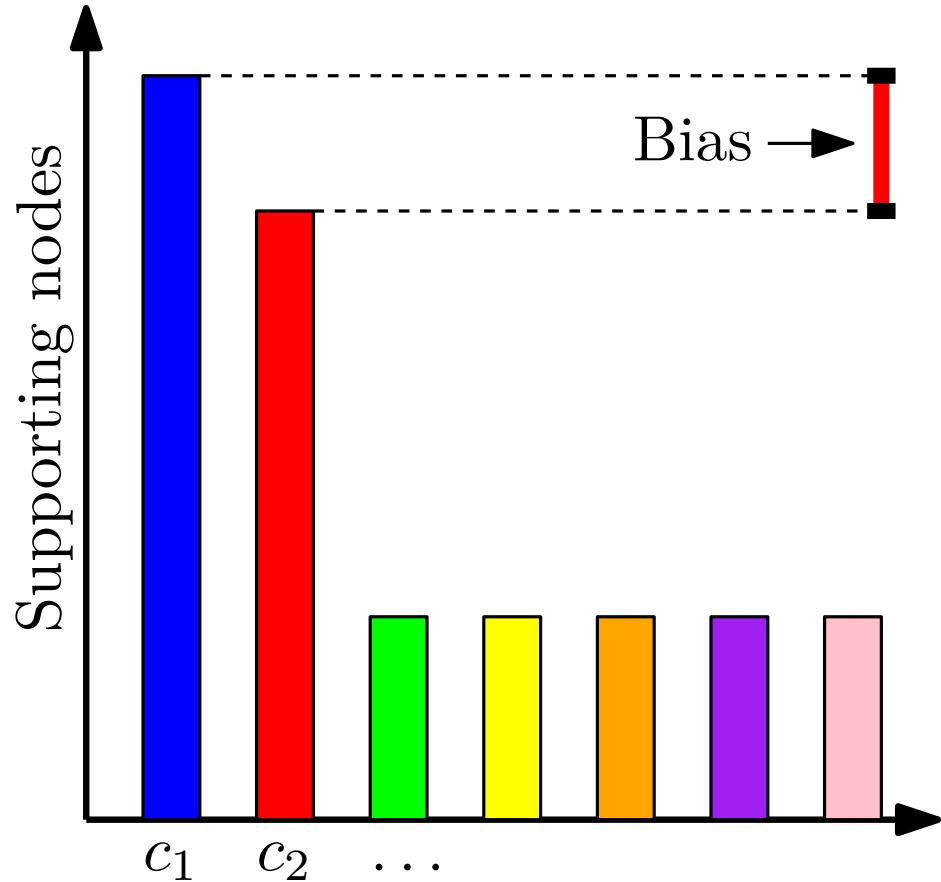
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Convergence of 3-Majority [SPAA '14]

Thm. From any configuration with $k < \sqrt[3]{n}$ colors, such that

$$s \geq 22\sqrt{2kn \log n},$$

the 3-majority protocol converges to the majority opinion in $O(2k \log n)$ rounds w.h.p., even in the presence of a $O(\sqrt{n})$ -bounded dynamic adversary.

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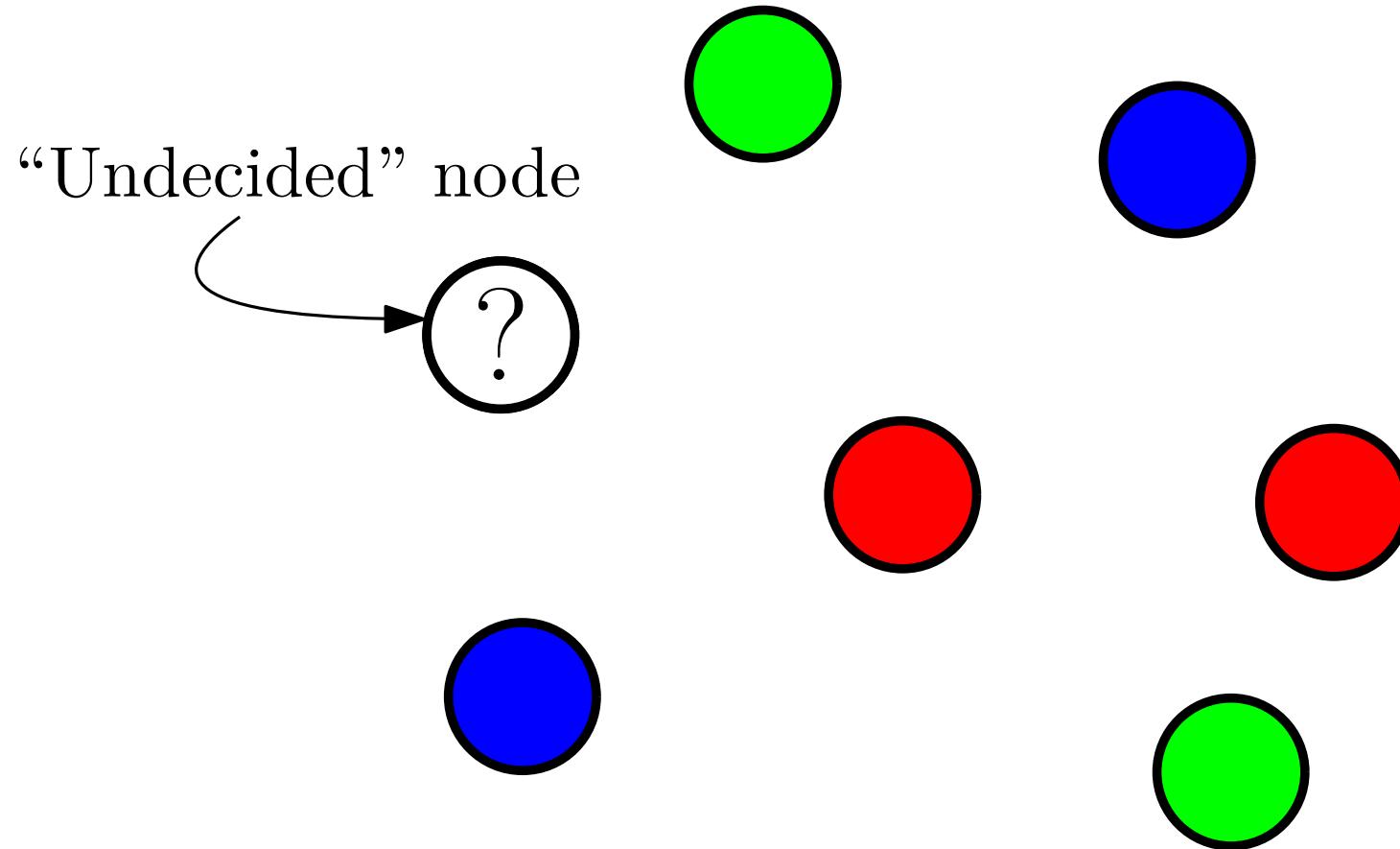
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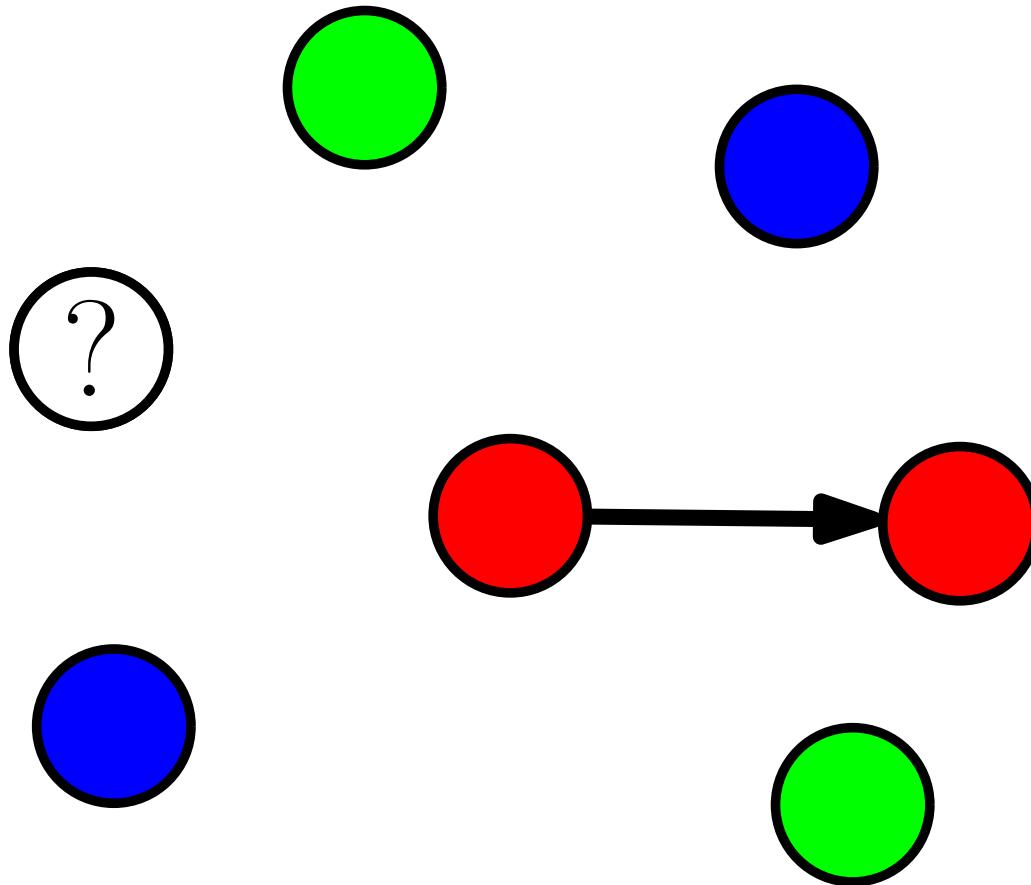
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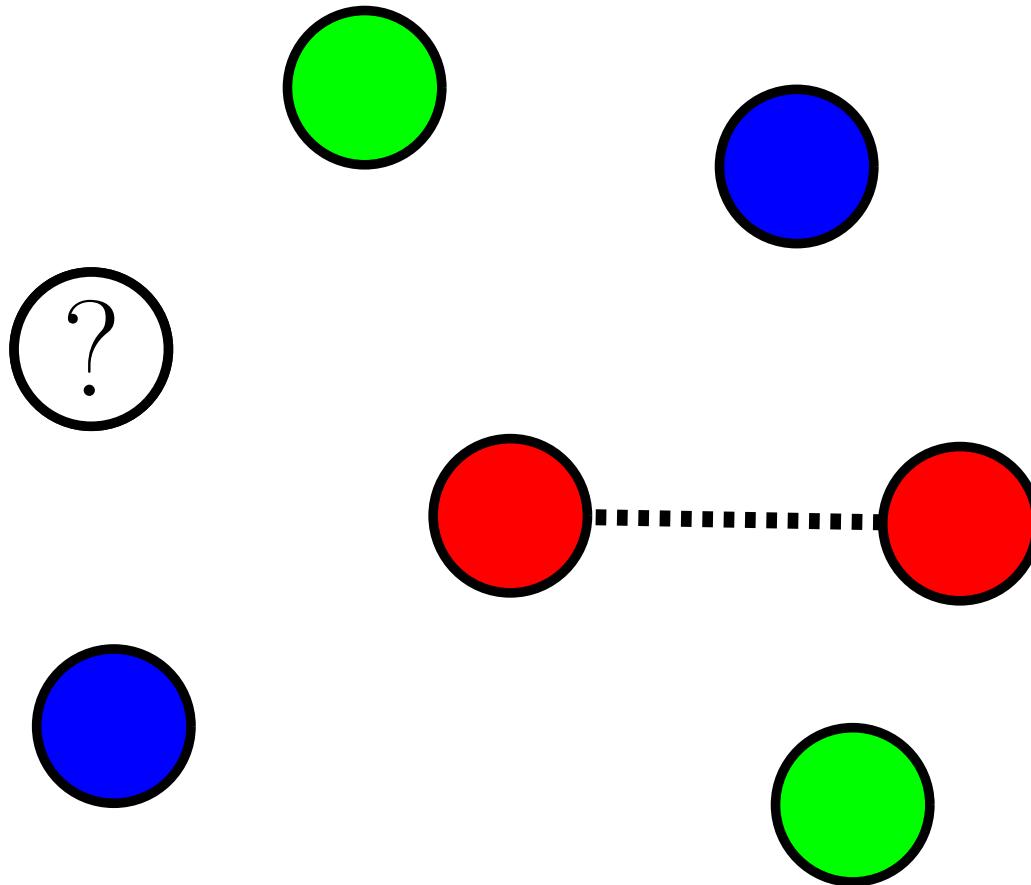
The Undecided-State Protocol



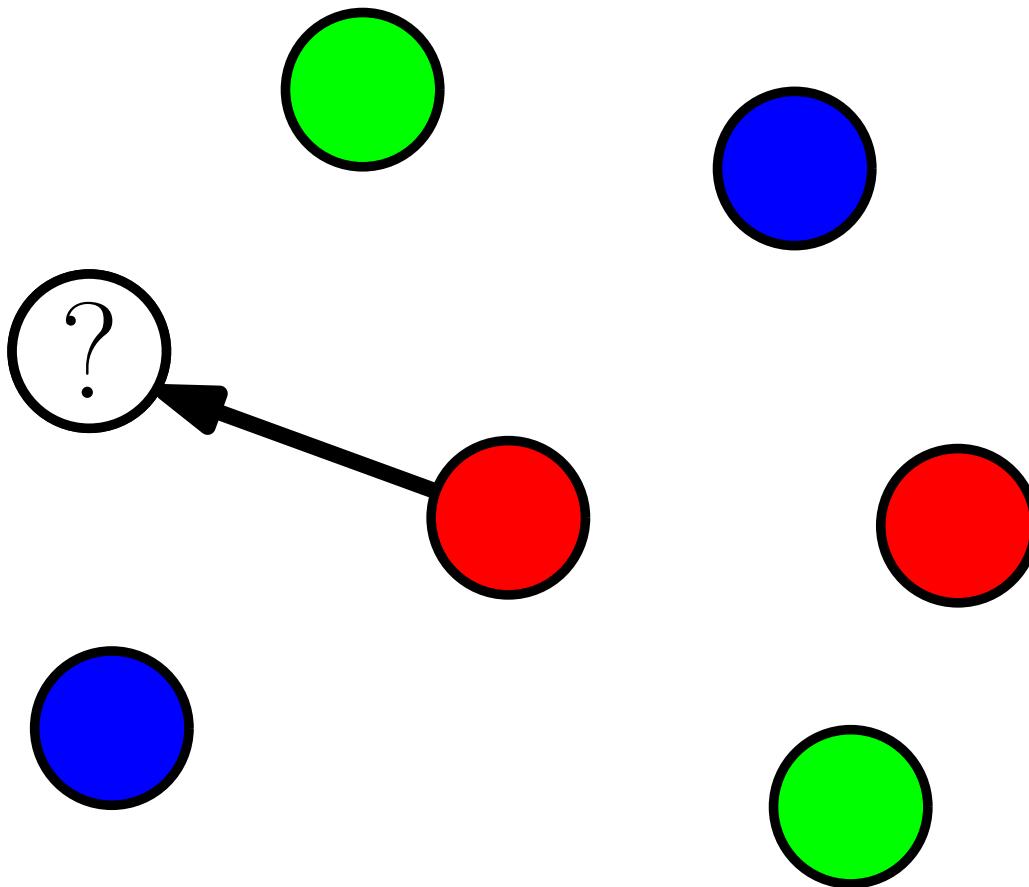
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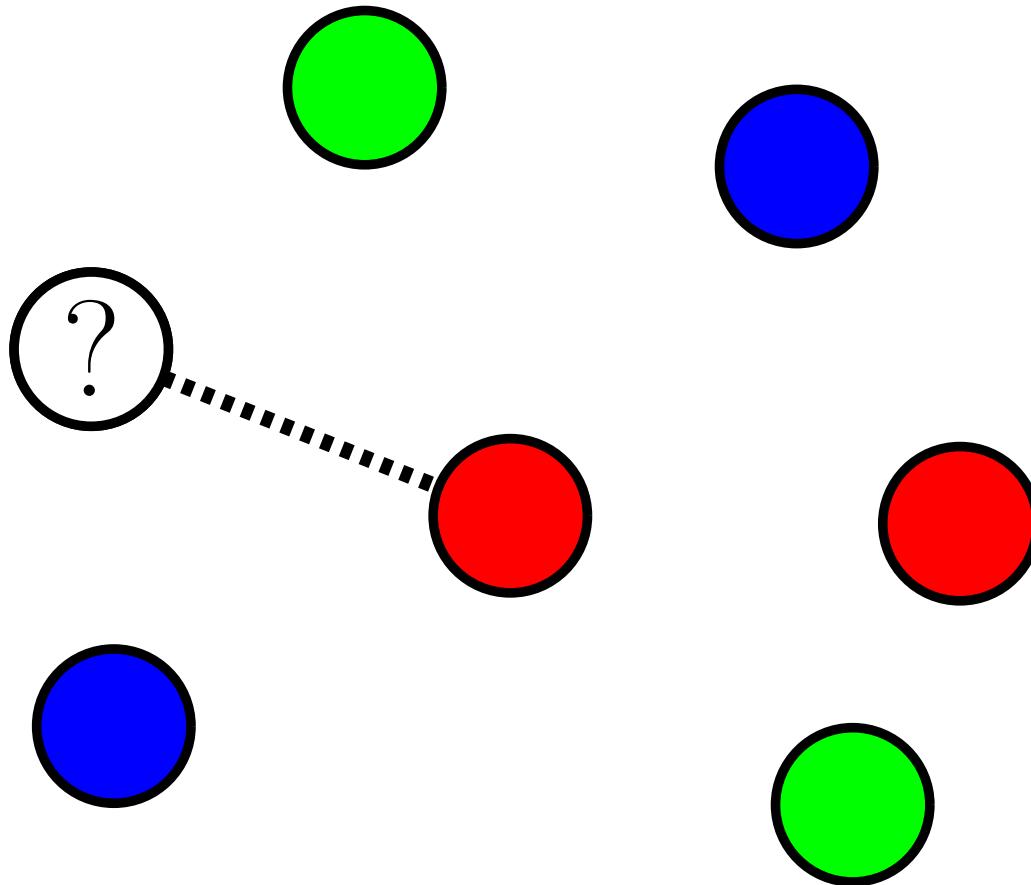
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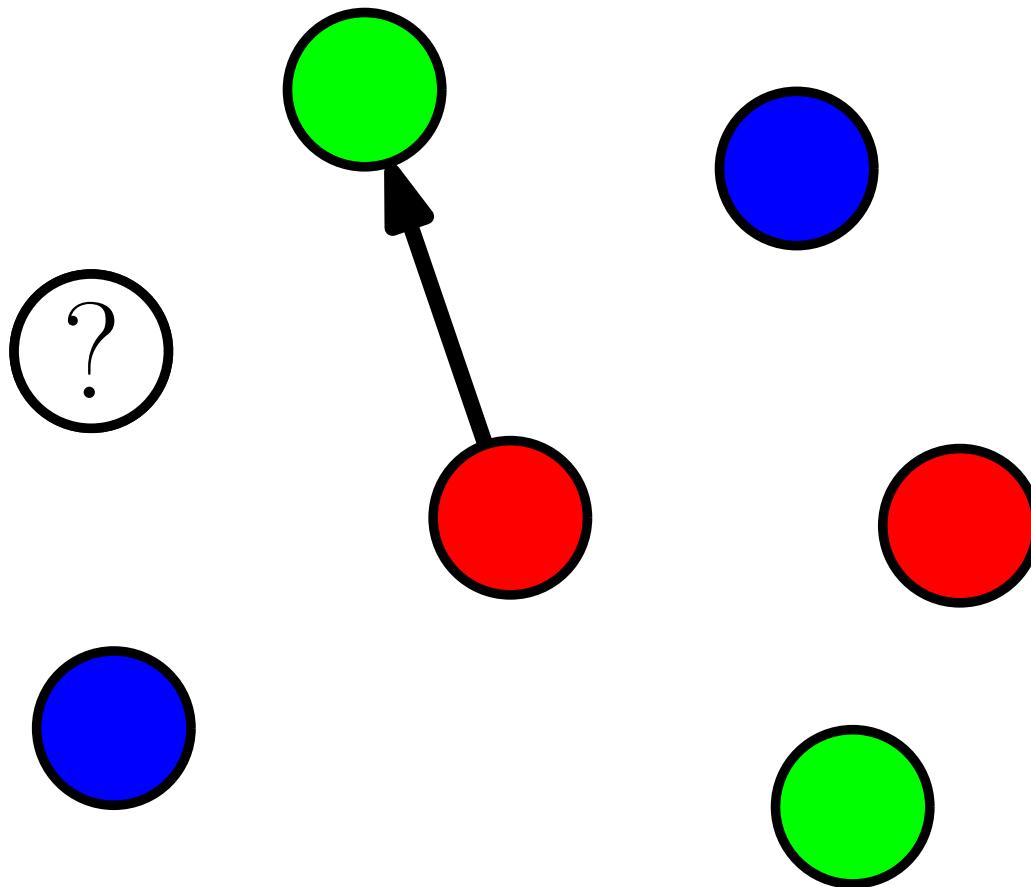
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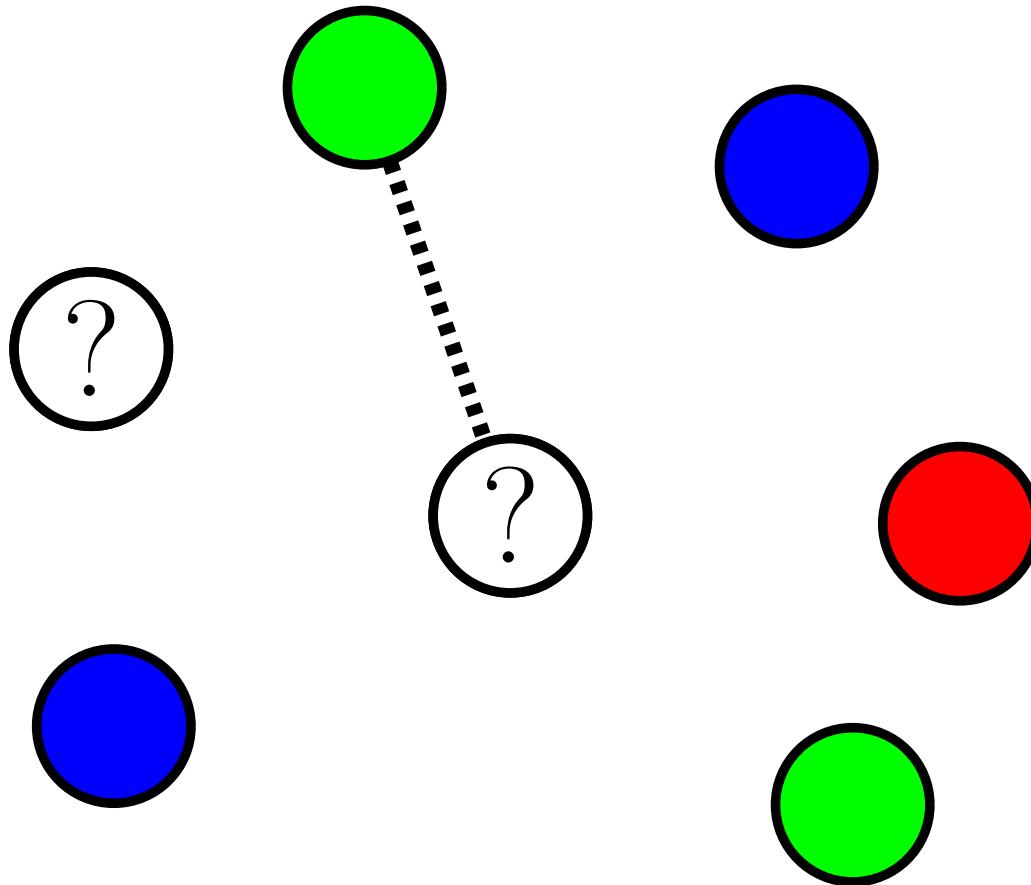
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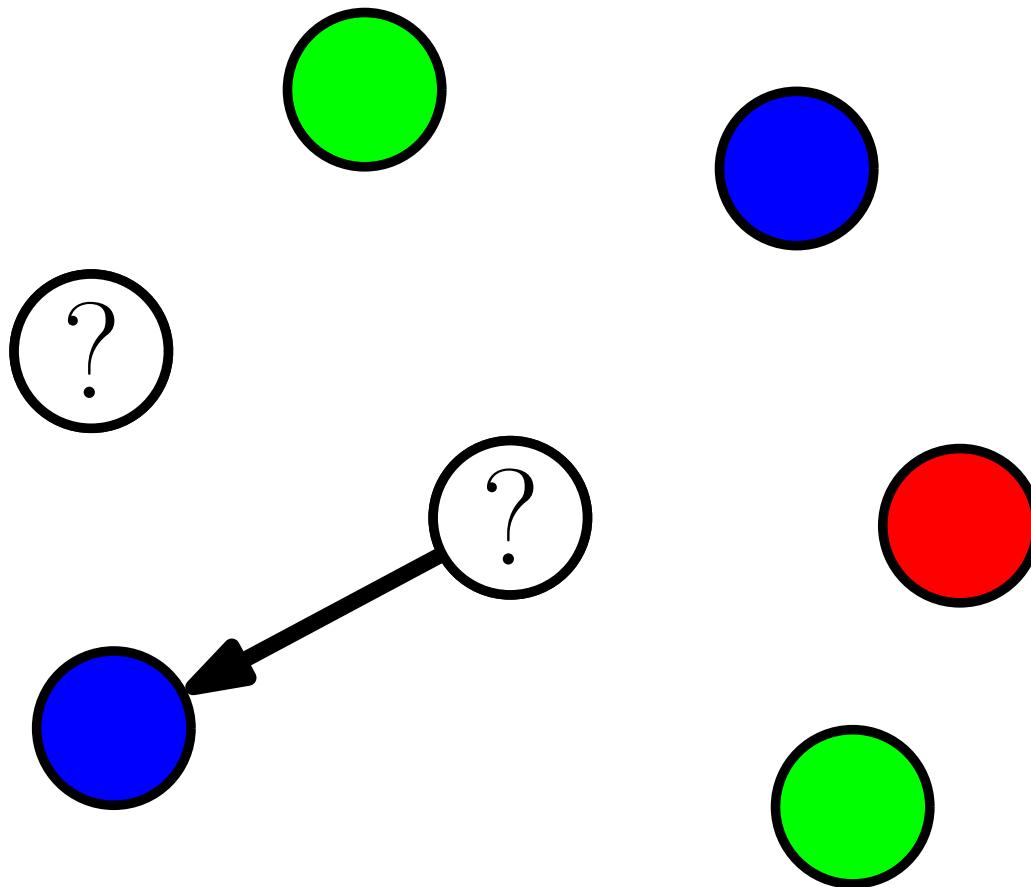
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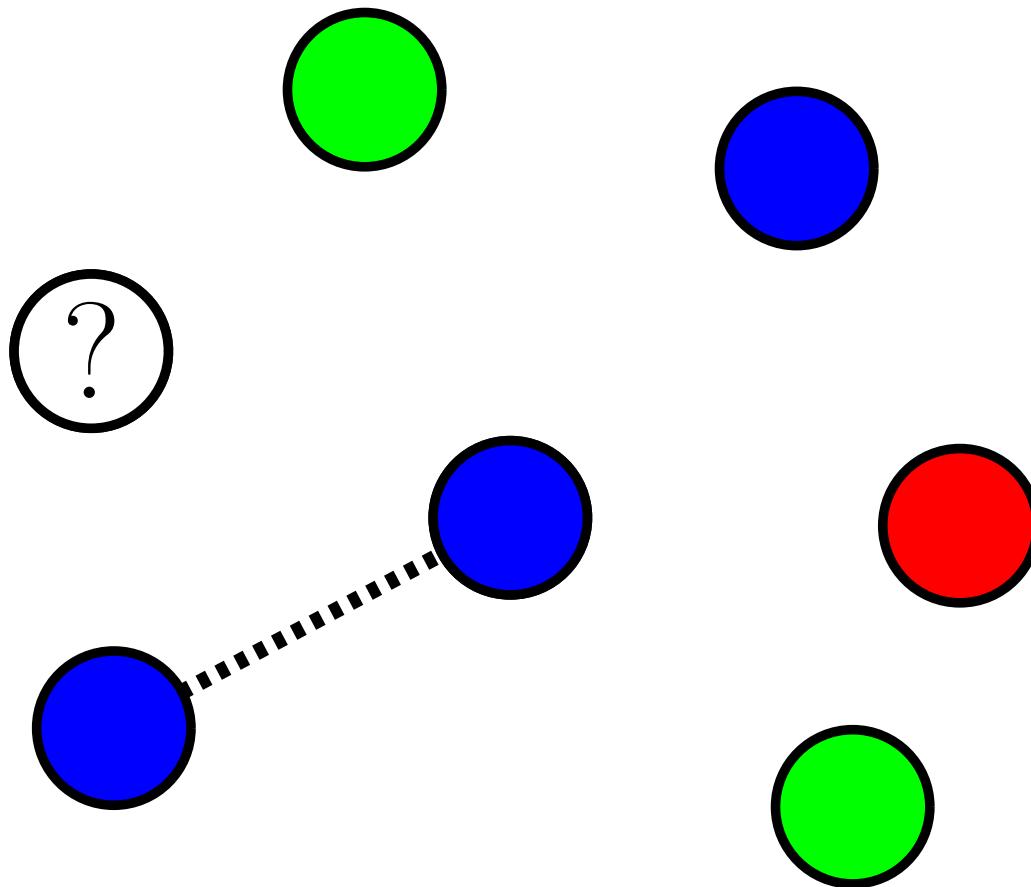
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The Monochromatic Distance

$c_i^{(t)} := \# \text{ nodes with color } i,$ $\mathbf{c}^{(t)} := \text{configuration at time } t.$

$$\text{md}(\mathbf{c}^{(0)}) := \sum_{i=1}^k \left(\frac{c_i^{(0)}}{c_1^{(0)}} \right)^2 = 1 + \mathcal{D} \left(\begin{array}{c} \text{Bar chart showing configuration at time 0: } \\ \text{A horizontal axis with vertical bars of different colors (blue, red, yellow, green, orange) of varying heights.} \end{array}, \begin{array}{c} \text{Bar chart showing configuration at time t: } \\ \text{A horizontal axis with a single tall blue bar.} \end{array} \right)$$

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$$1 \leq \text{md} \left(\begin{array}{c} \text{Bar chart of } \mathbf{c}^{(0)} \\ \text{with } k \text{ bars} \end{array} \right) \ll \text{md} \left(\begin{array}{c} \text{Bar chart of } \mathbf{c}^{(0)} \\ \text{with } k \text{ bars} \end{array} \right) \leq k$$

Convergence of the Undecided-State [SODA '15]

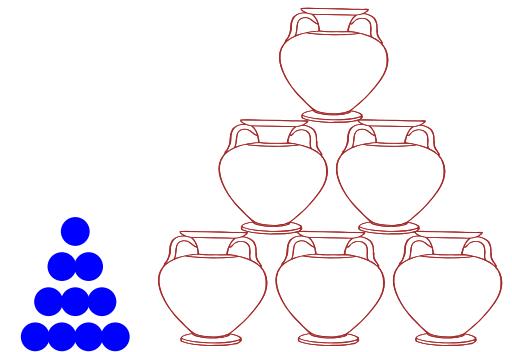
Theorem

If $k = O\left((n/\log n)^{1/3}\right)$ and $c_1 \geq (1 + \epsilon) \cdot c_2$, then w.h.p. the Undecided-State Dynamics reaches plurality consensus in $O\left(\text{md}(\mathbf{c}^{(0)}) \cdot \log n\right)$ rounds.

Thank You!

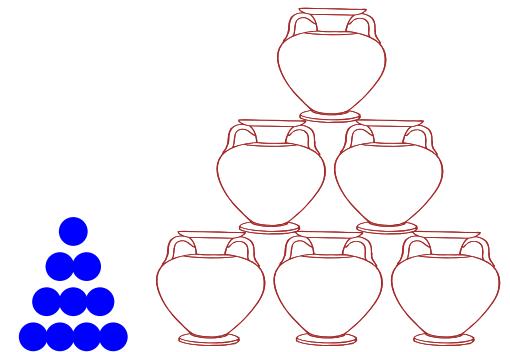
Other Stuff with My Group

Probabilistic Self-Stabilization:
Repeated Balls into Bins [SPAA '15]



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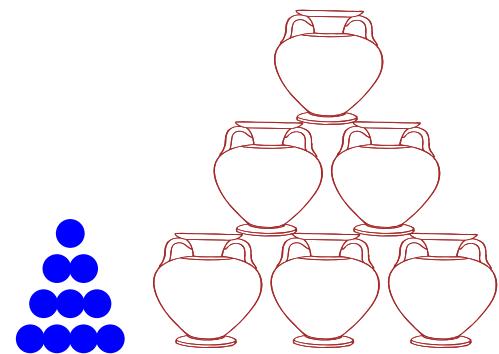
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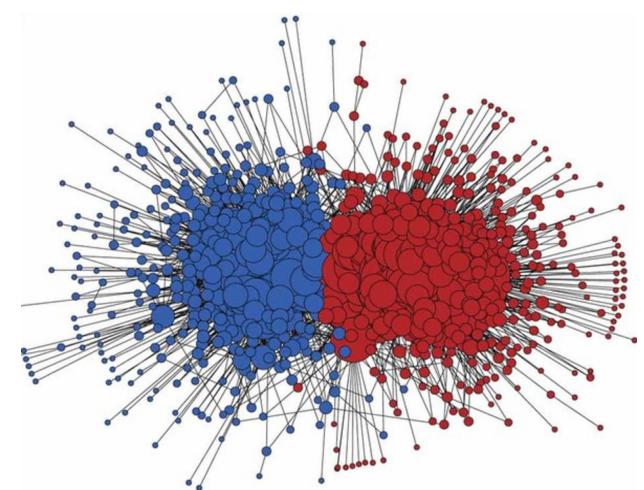
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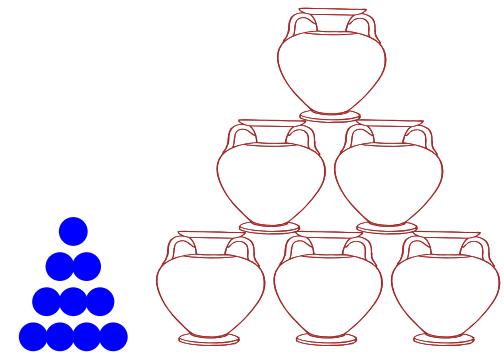
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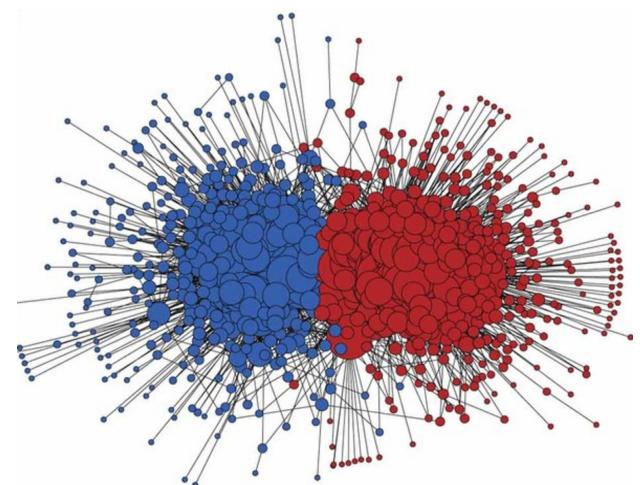
Distributed Community Detection
in Stochastic Block Models
[TCS '15 + Coming soon]

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Jan-May at the Simons Institute

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With L. Gualà and S. Leucci:
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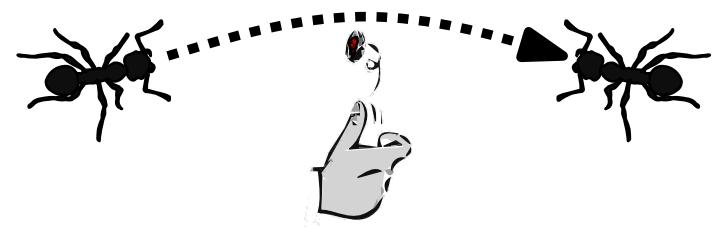
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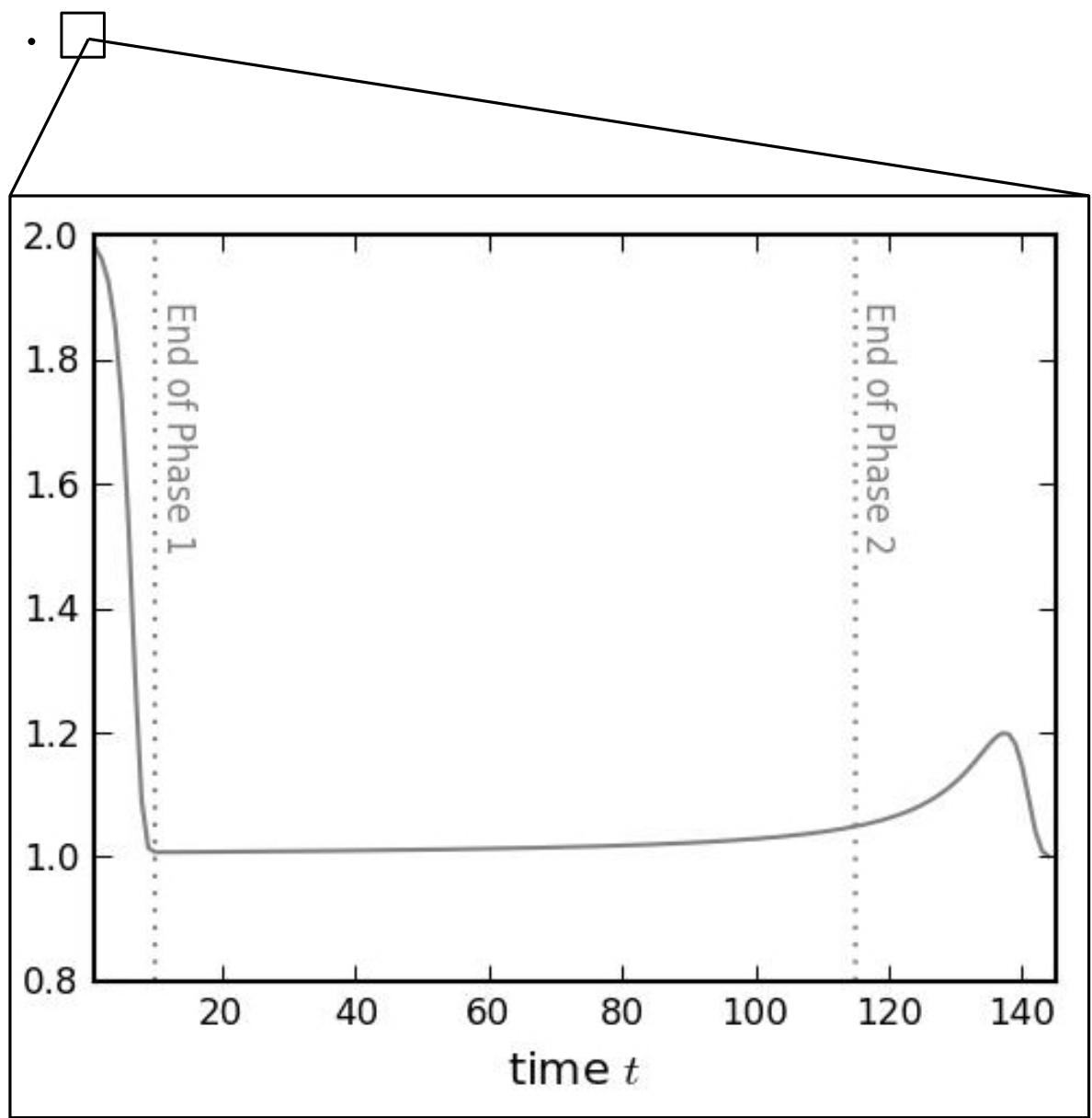
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With P. Fraigniaud: “Natural”
Consensus with Noisy
Communication [Coming soon]



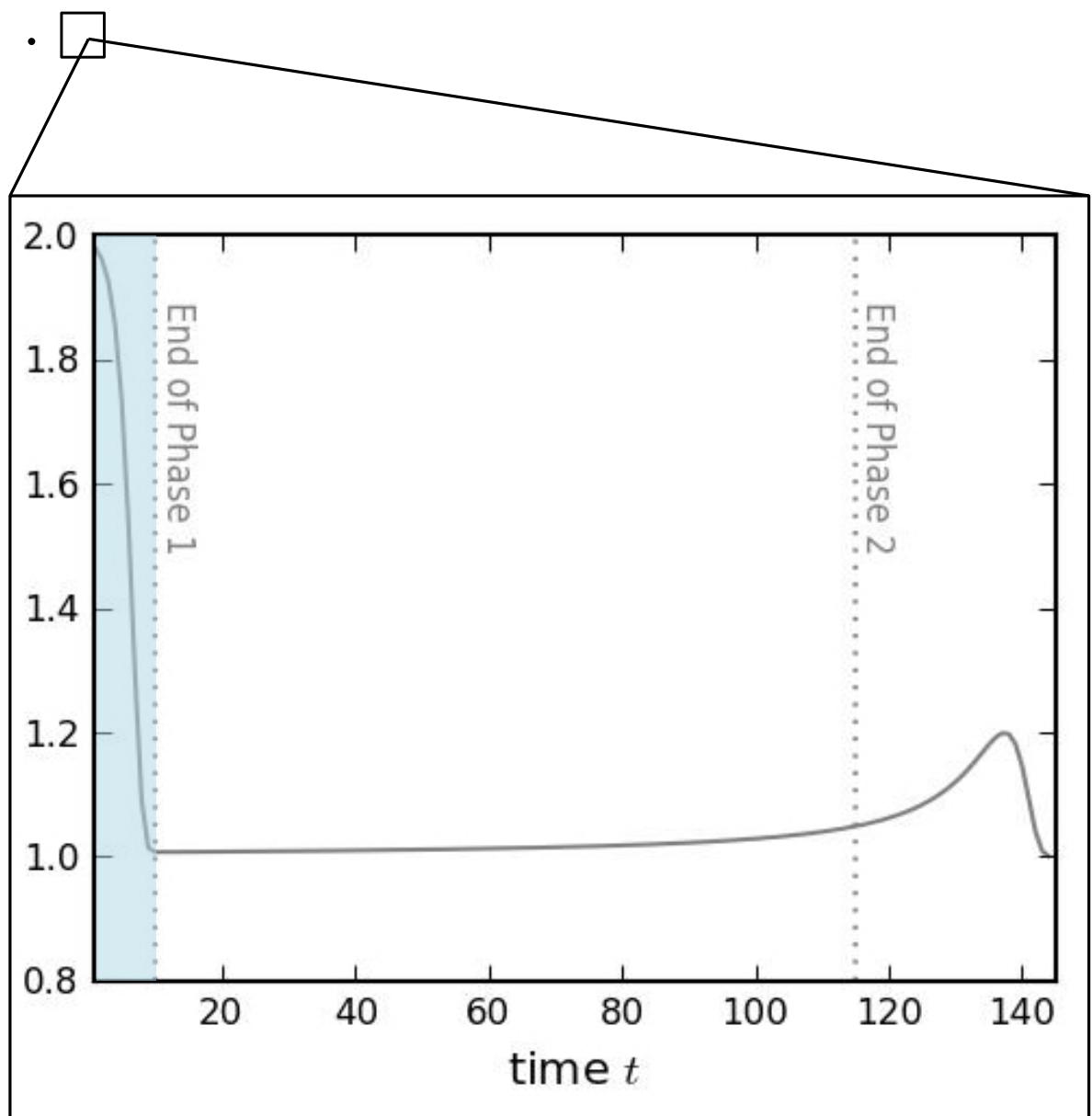
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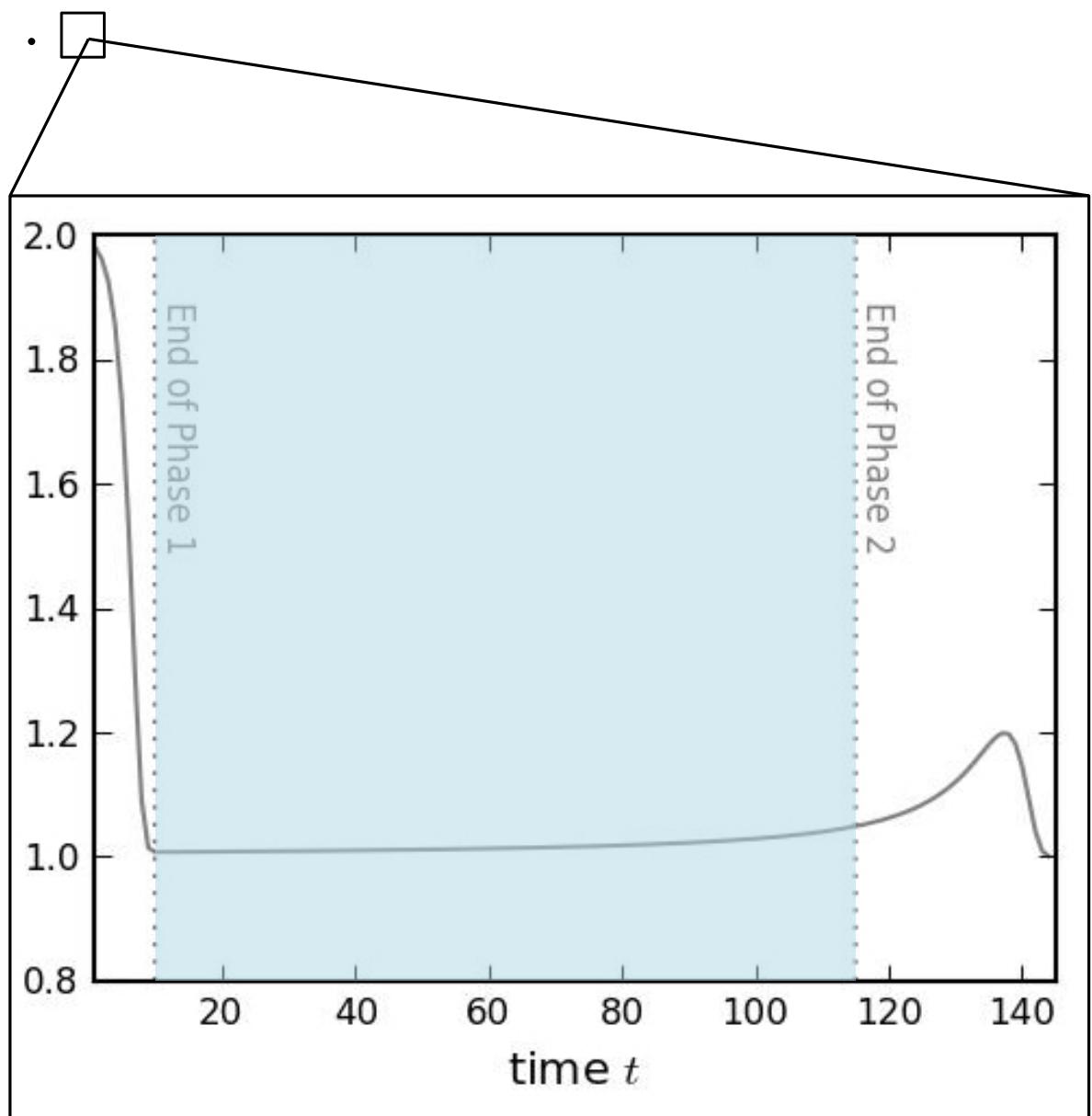
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