

# Gossip Algorithms for Majority Consensus

Emanuele Natale

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

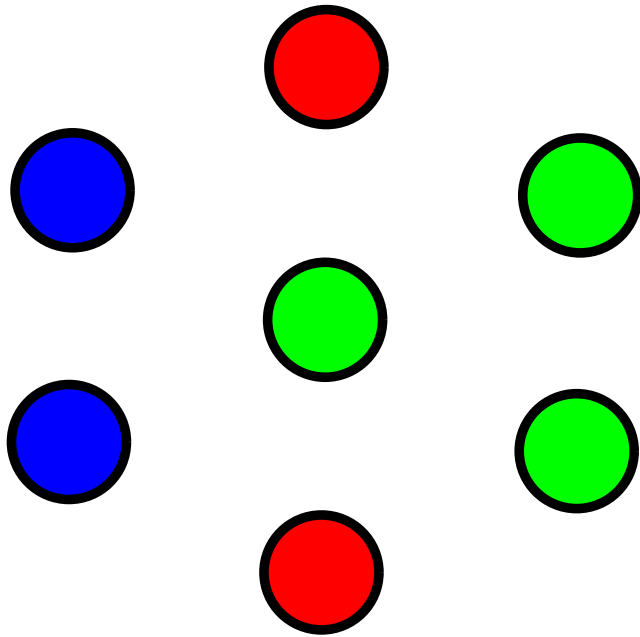
Research group: L. Becchetti, A. Clementi, F. Pasquale, R. Silvestri, (L. Trevisan) & me



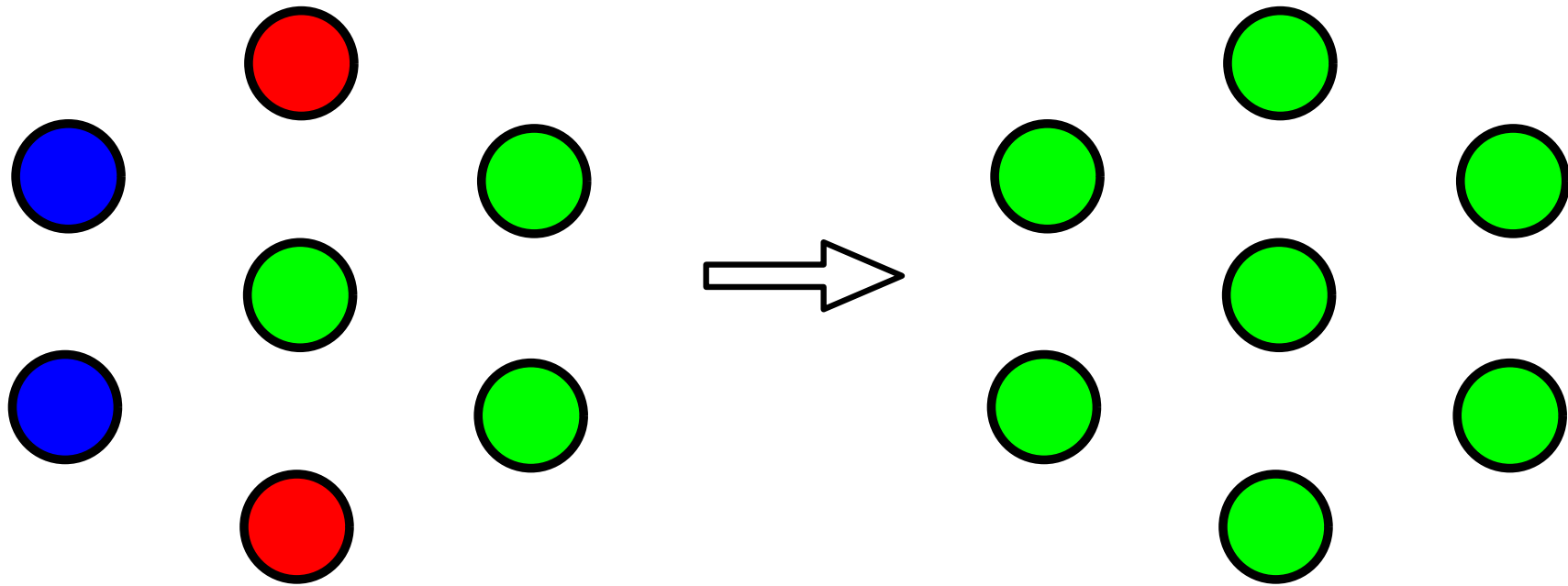
SAPIENZA  
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October 12, 2015

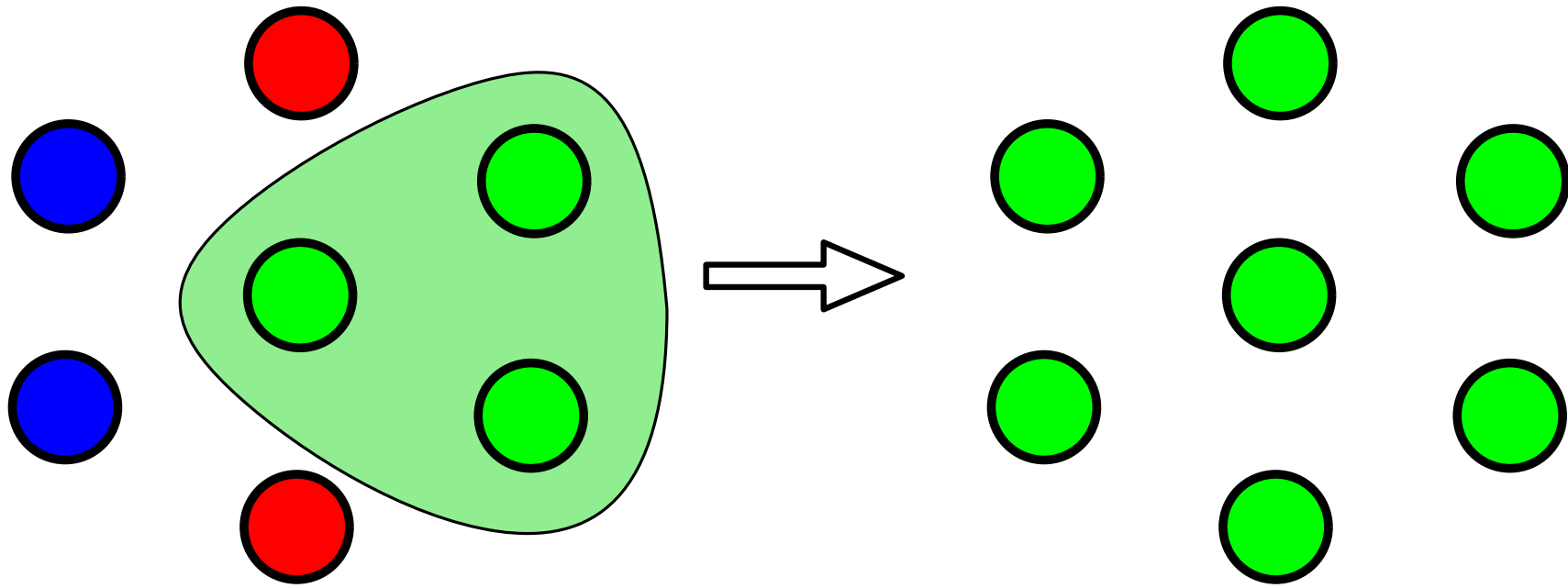
# The Majority Consensus Problem



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

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

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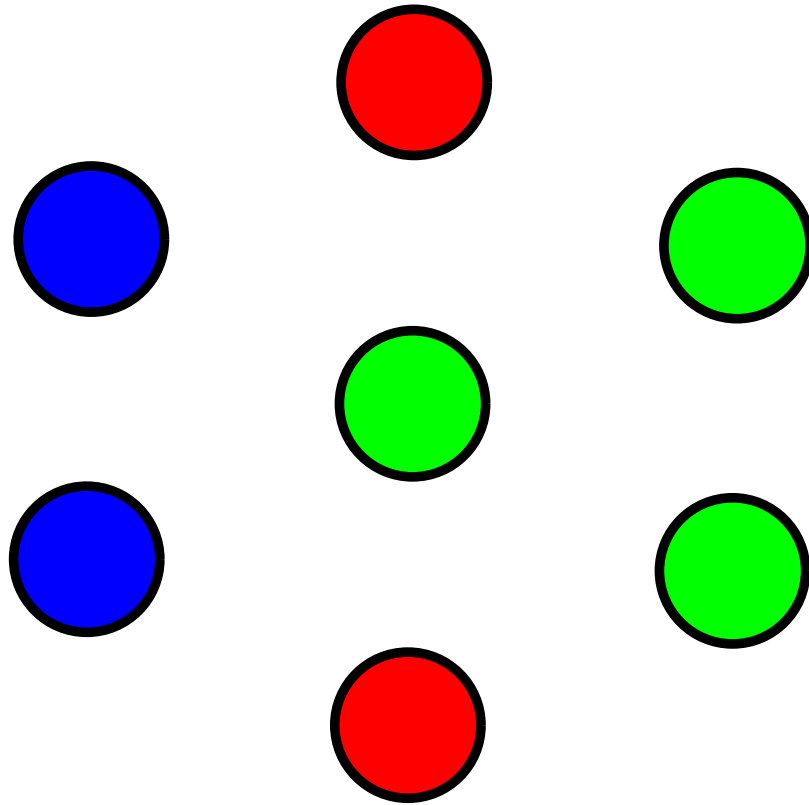
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Theoretical interest:

what can we do with minimal assumptions?

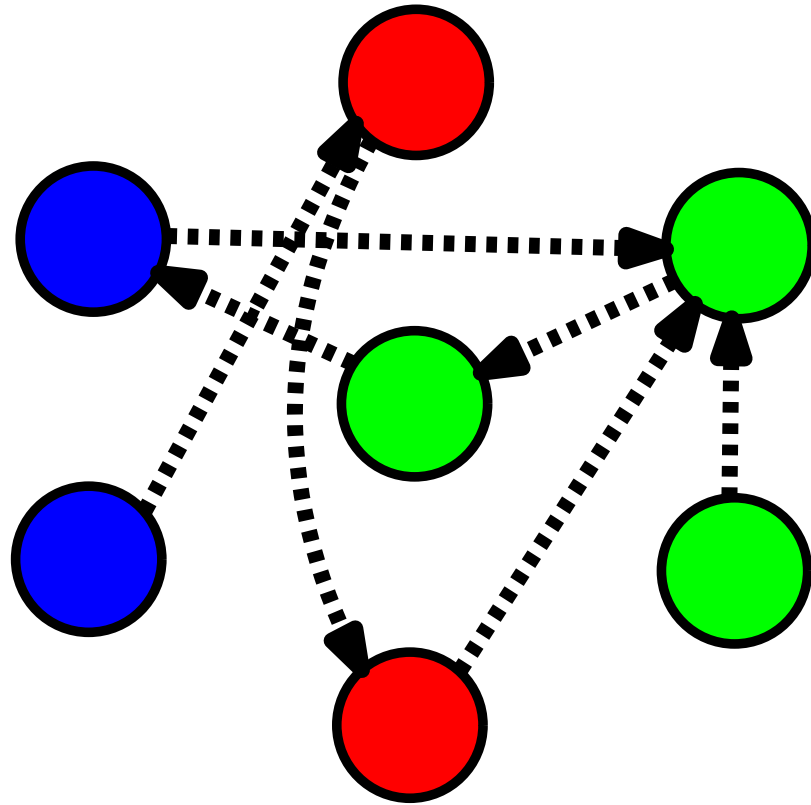
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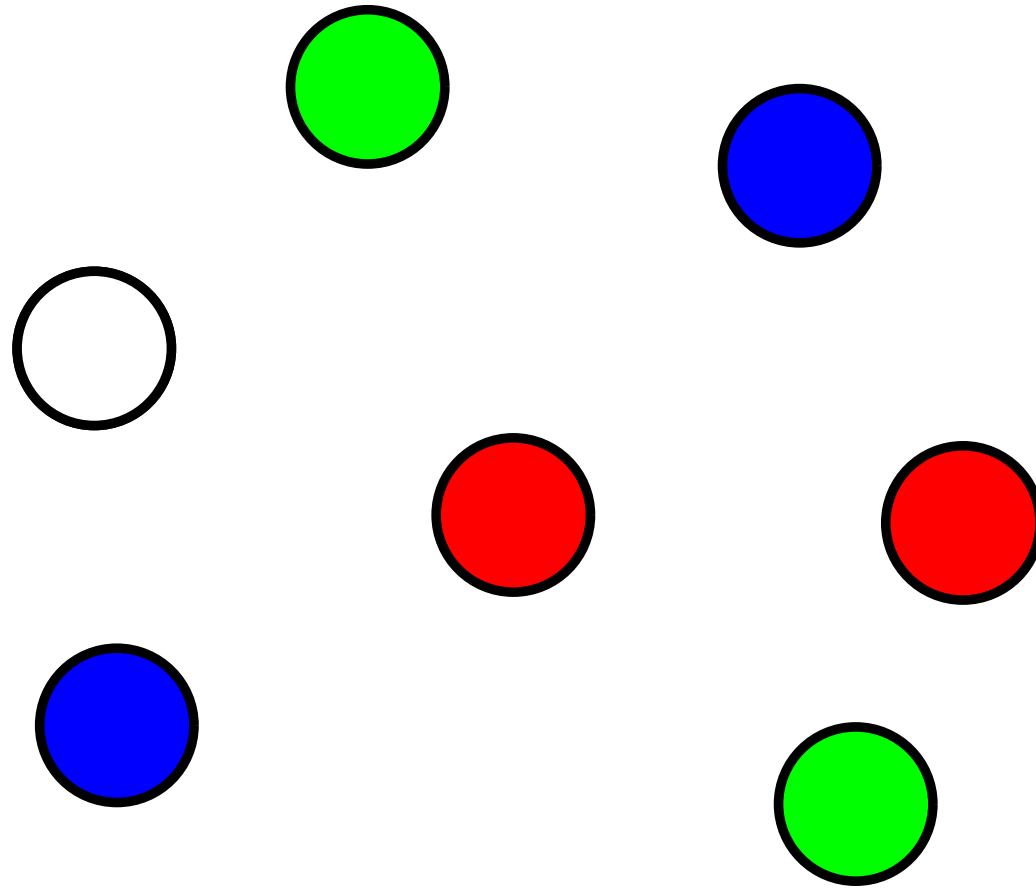
# (Some) Related Works

	Mem. & mess. size	# of colors	Time efficiency	Comm. Model
Kempe <sup>et al.</sup> FOCS '03	$O(k \log n)$	any	$O(\log n)$	<i>GOSSIP</i>
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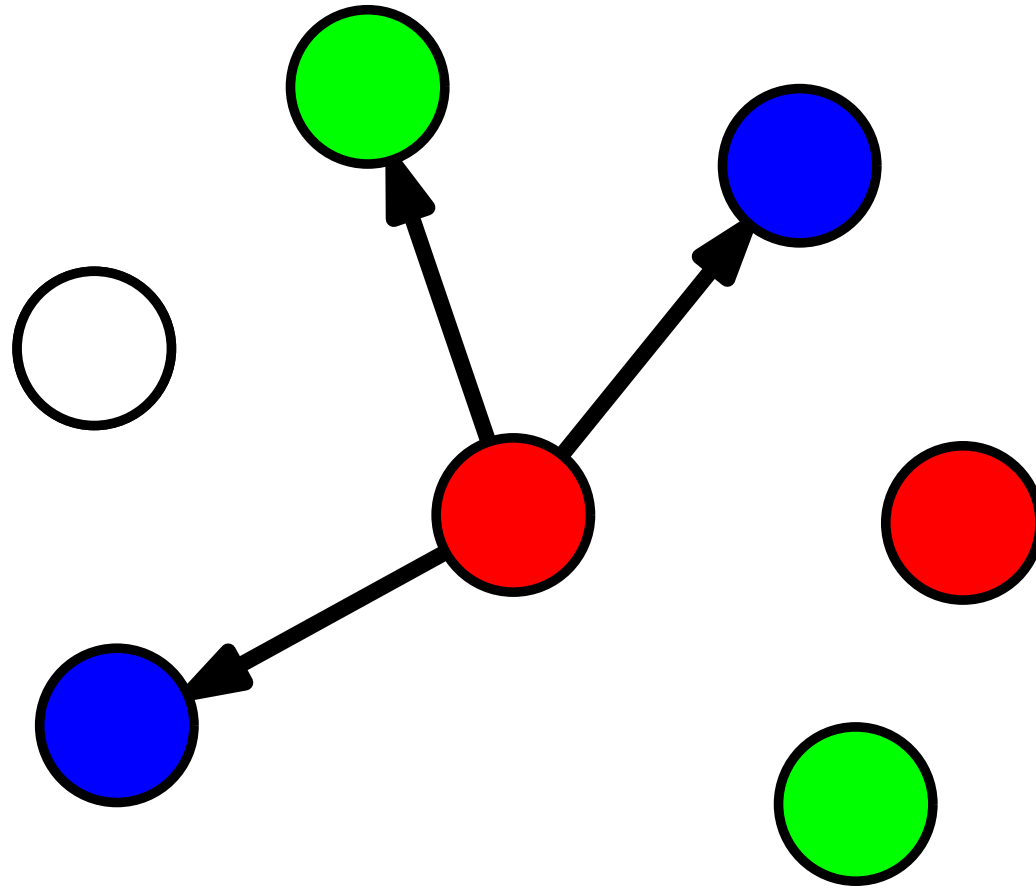
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# The 3-Majority Protocol

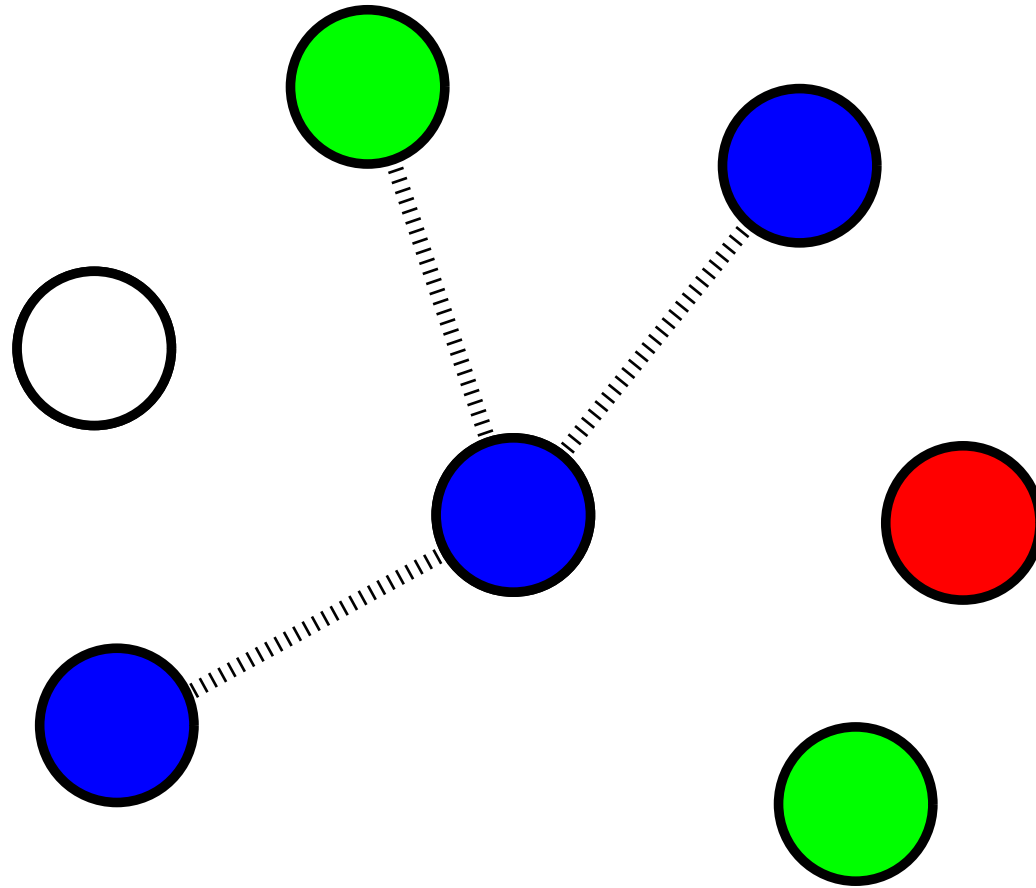


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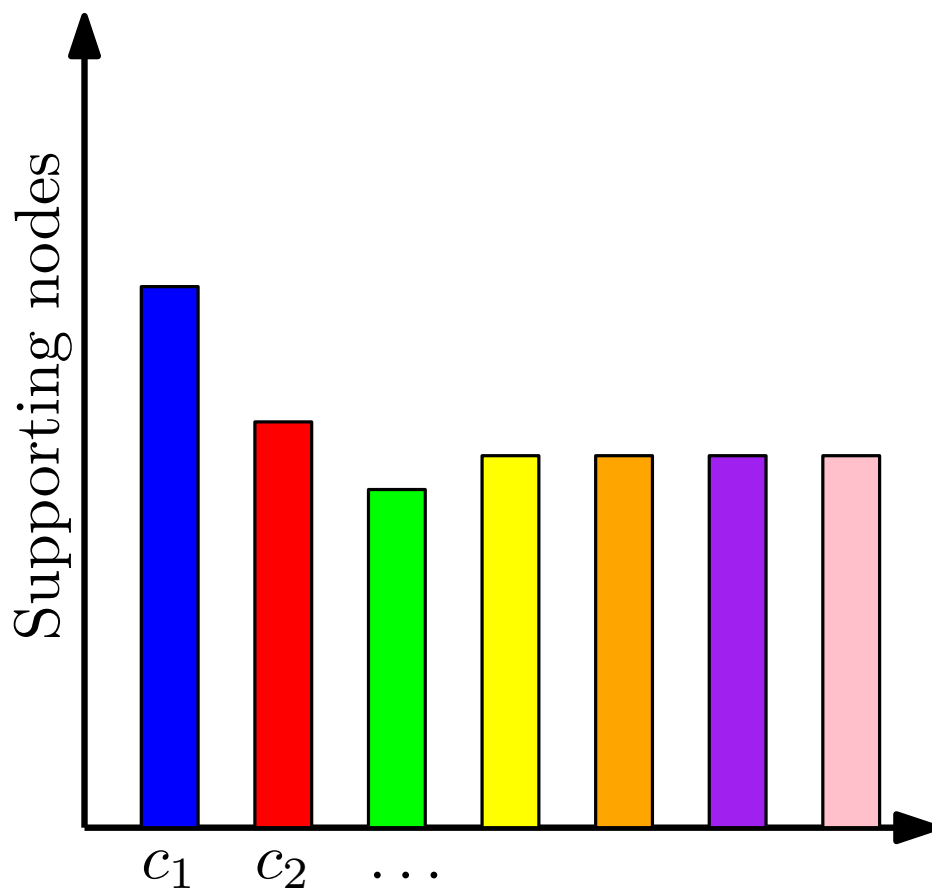
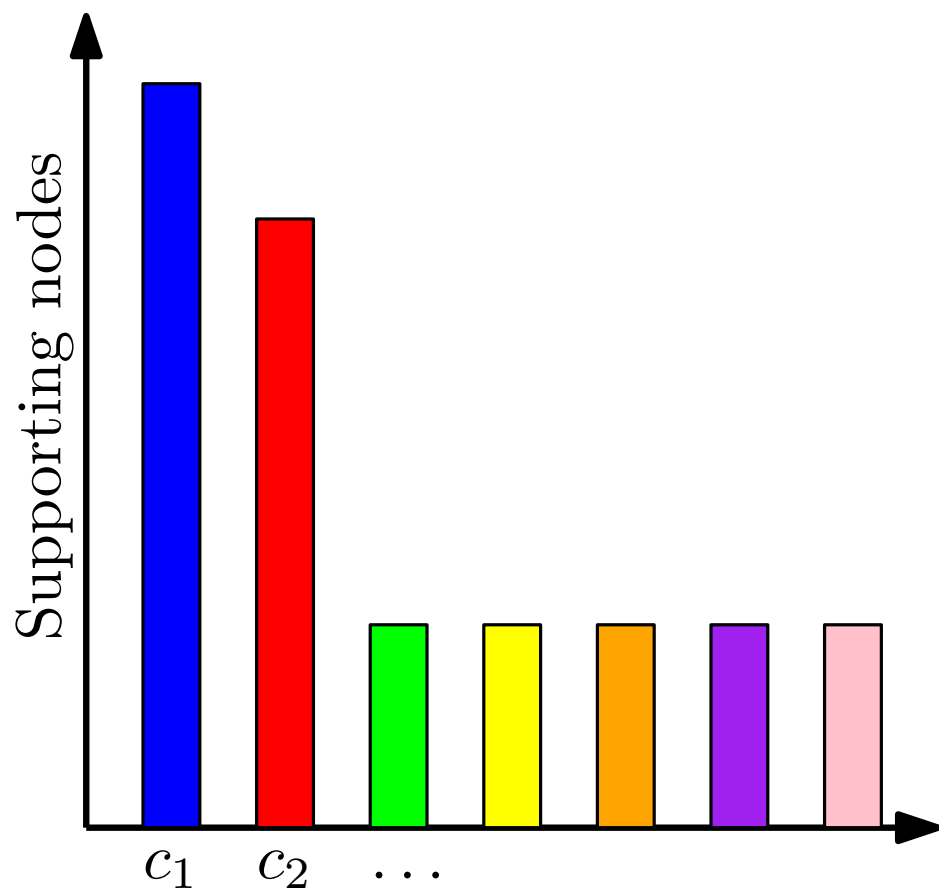
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# Key Parameter of 3-Majority

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

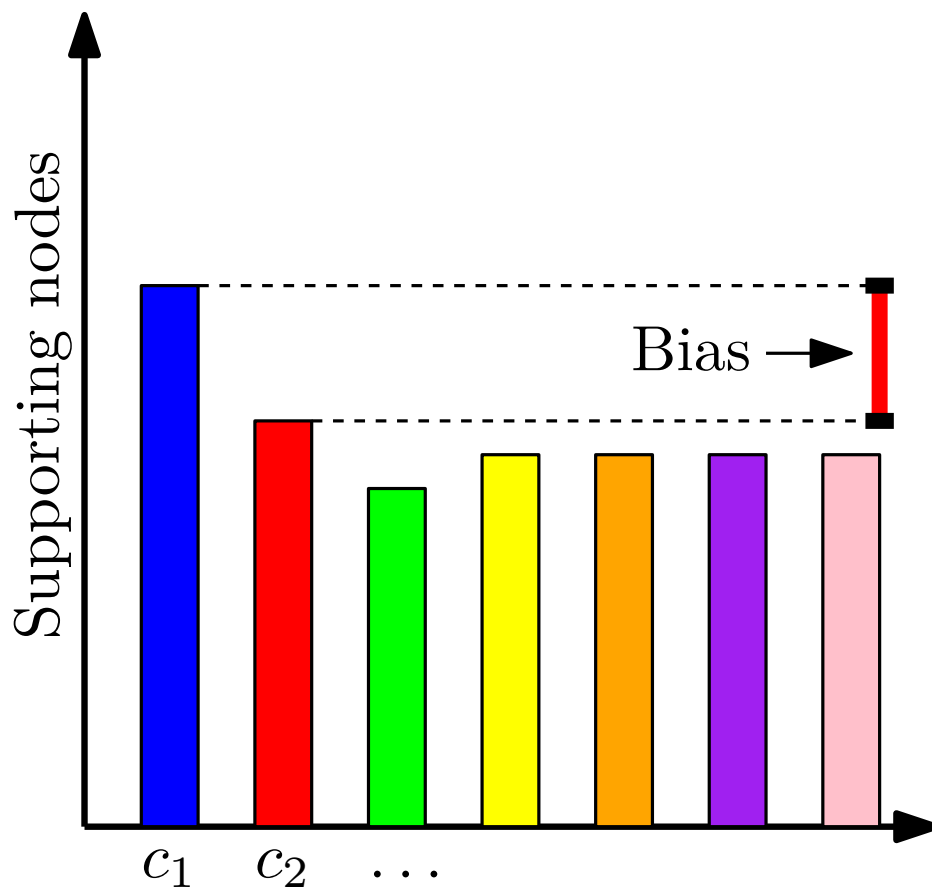
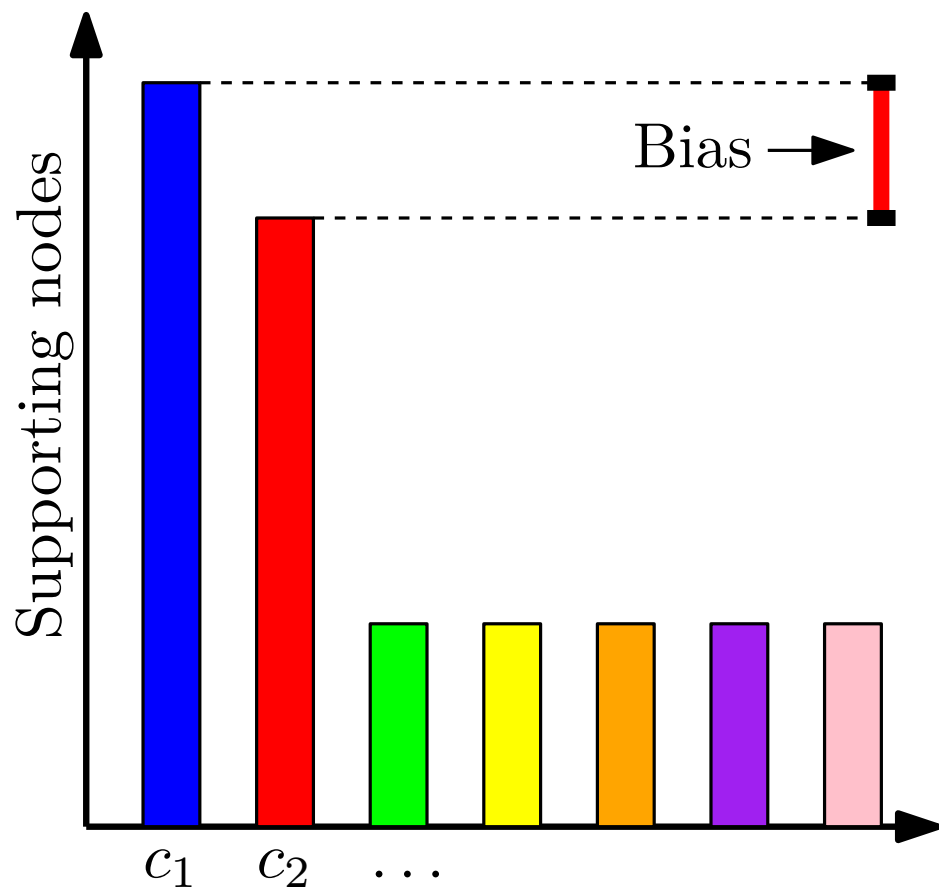
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# Convergence of 3-Majority [SPAA '14]

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

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

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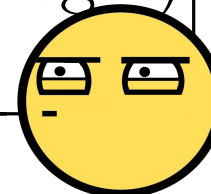
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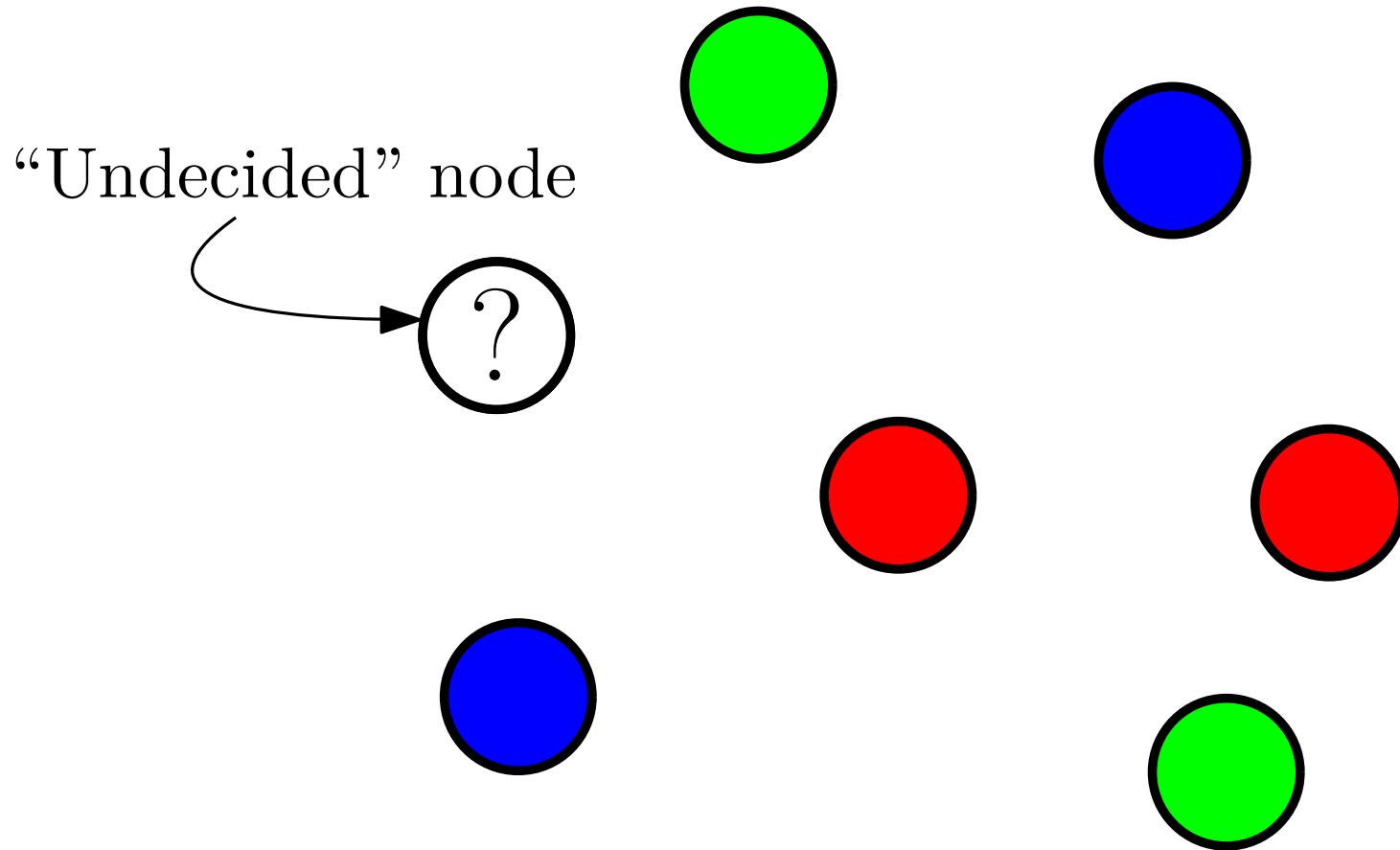
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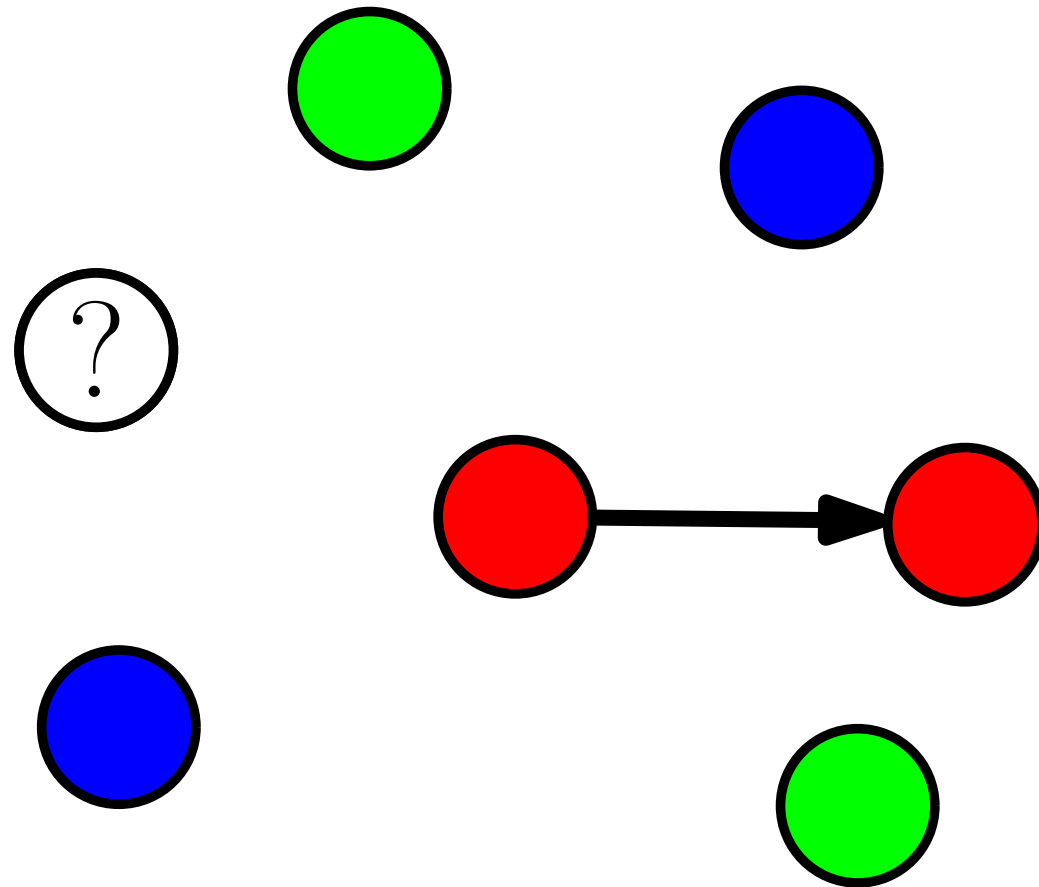


# The Undecided-State Protocol

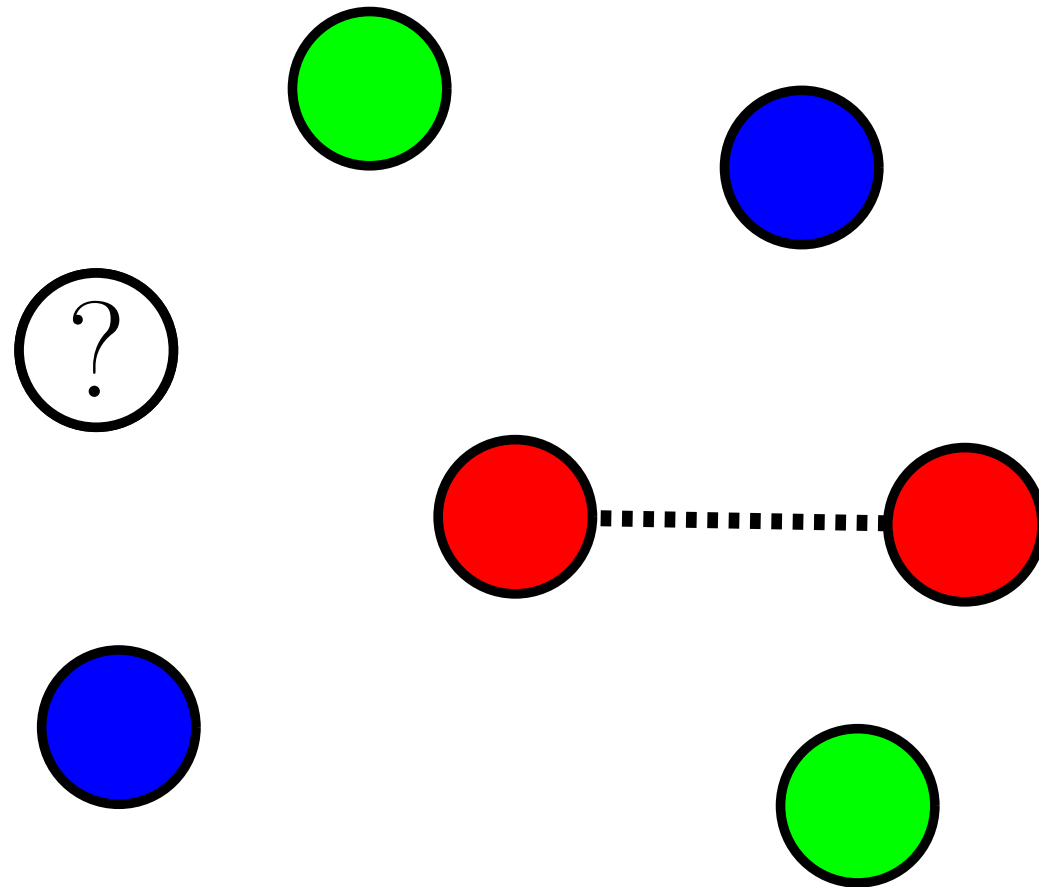




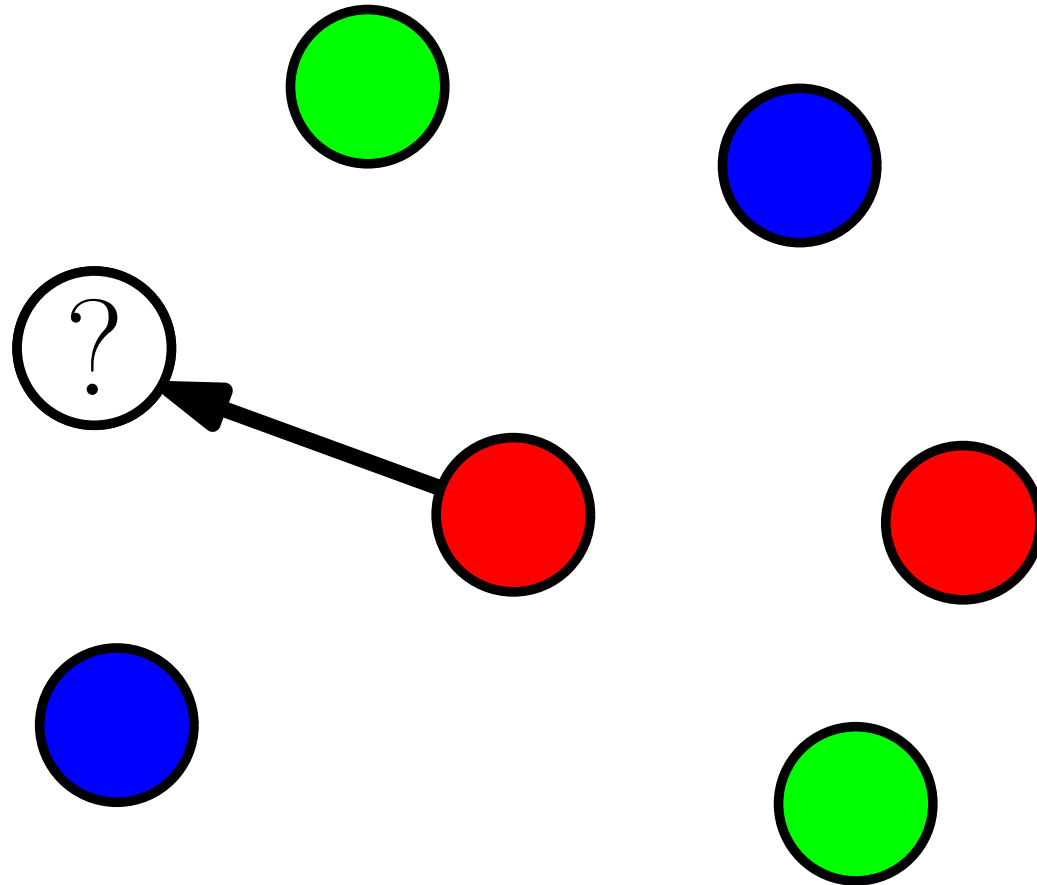
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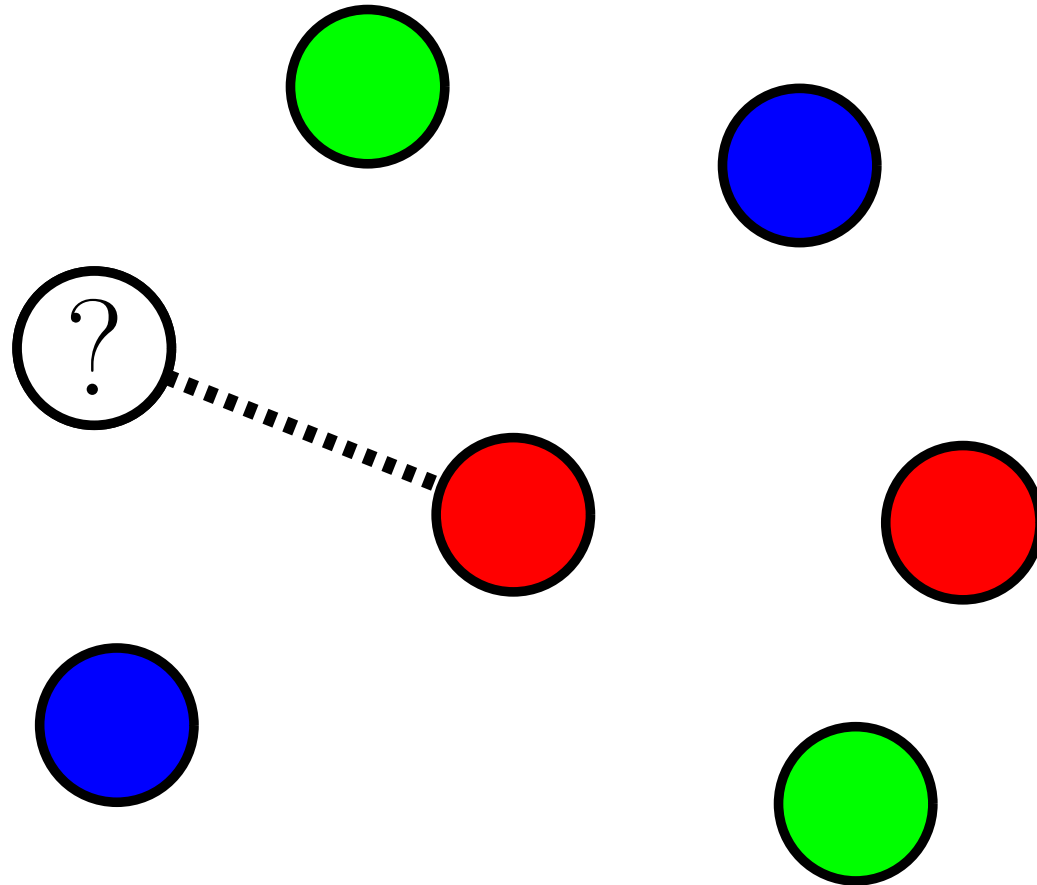
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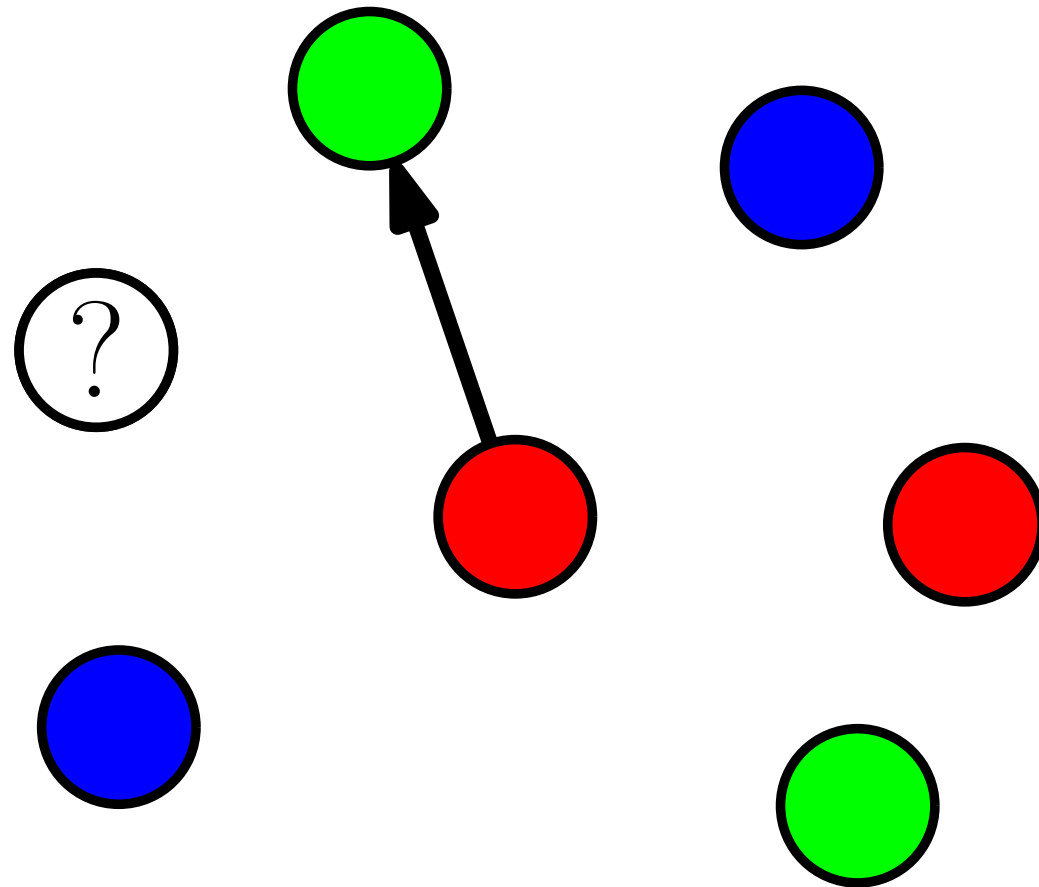
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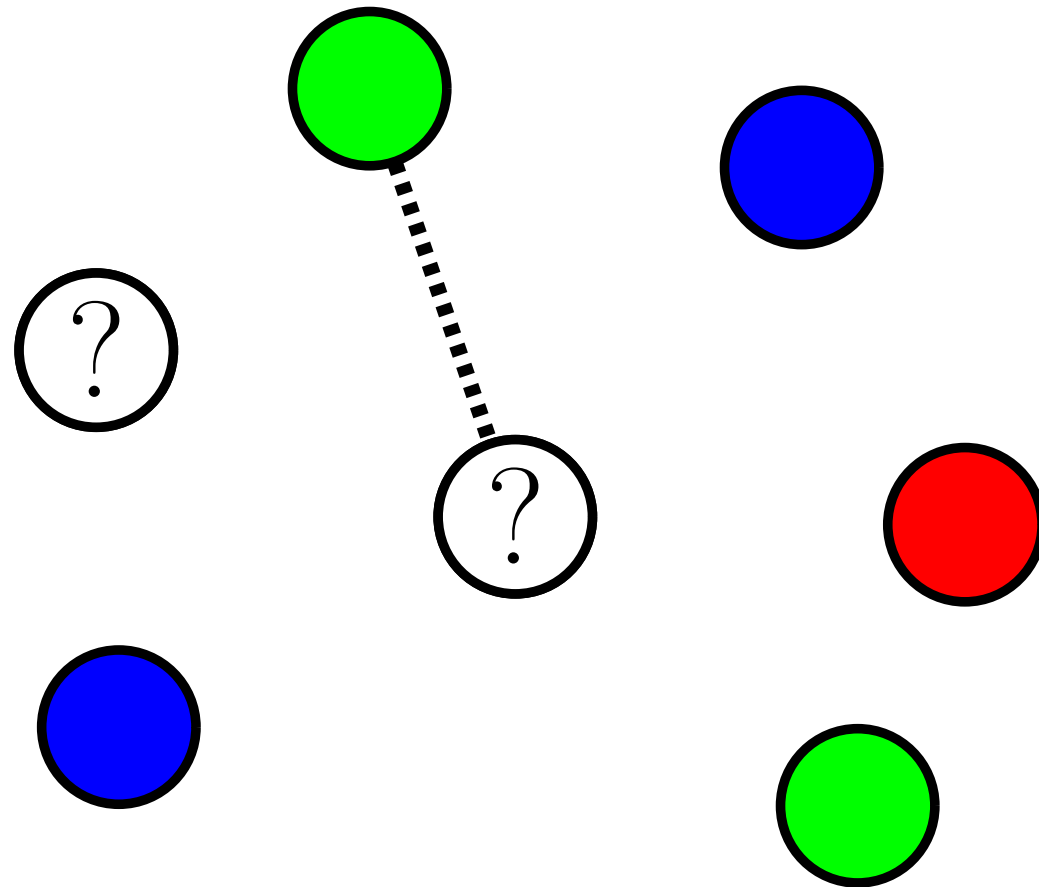
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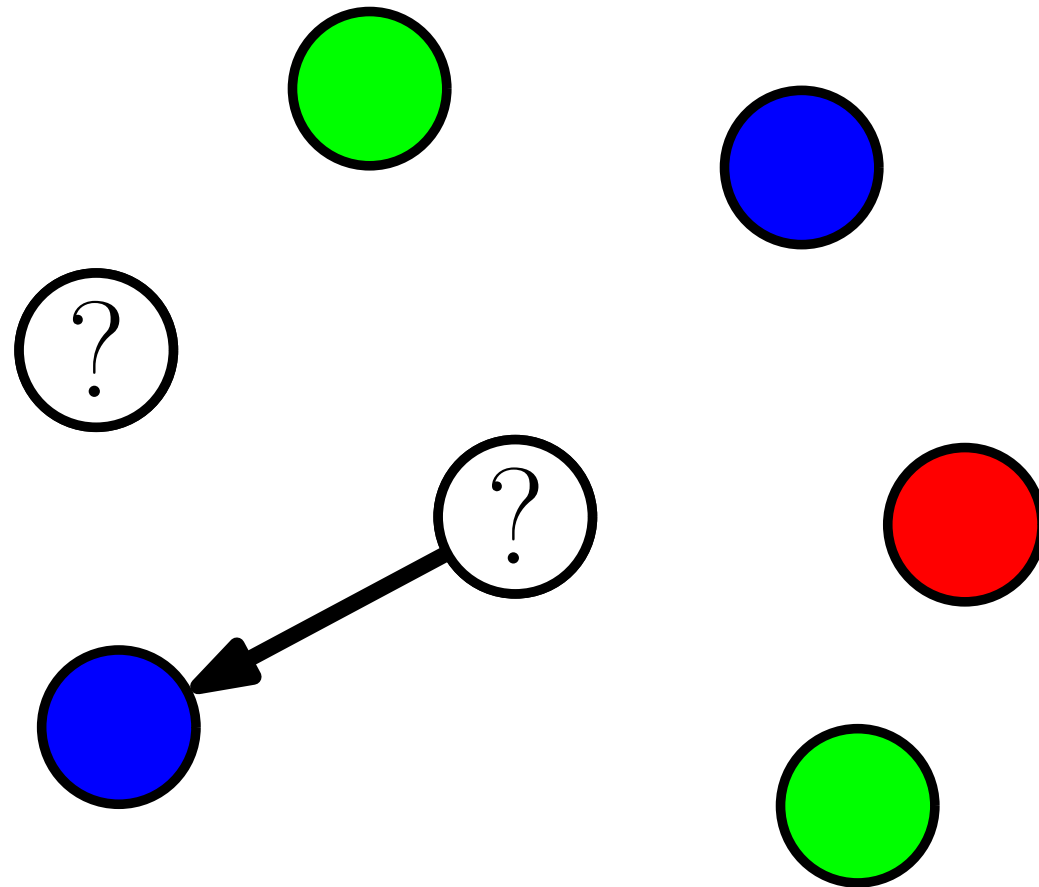
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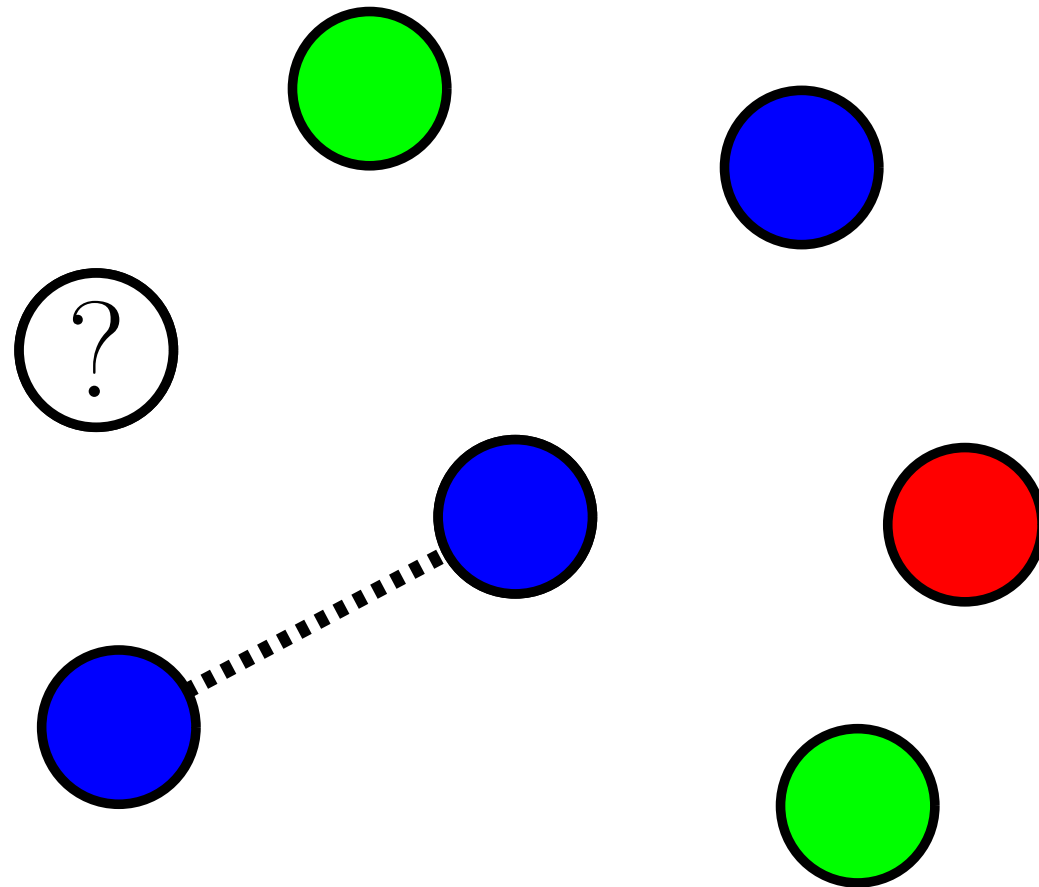
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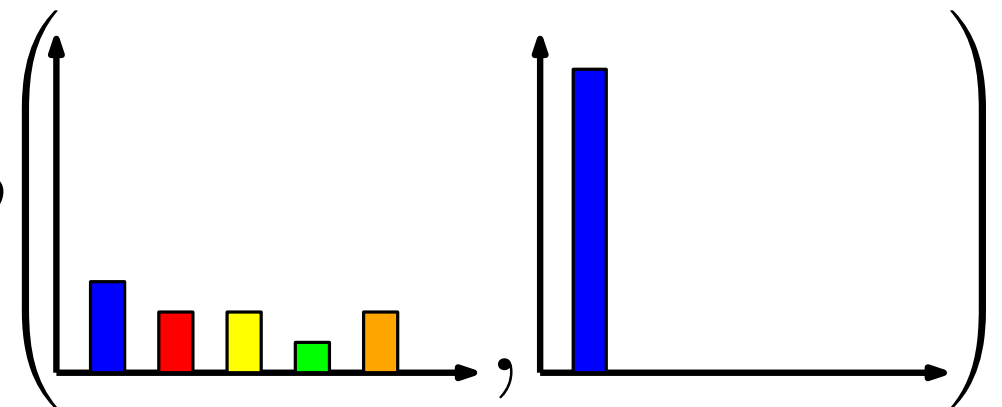
# The Monochromatic Distance

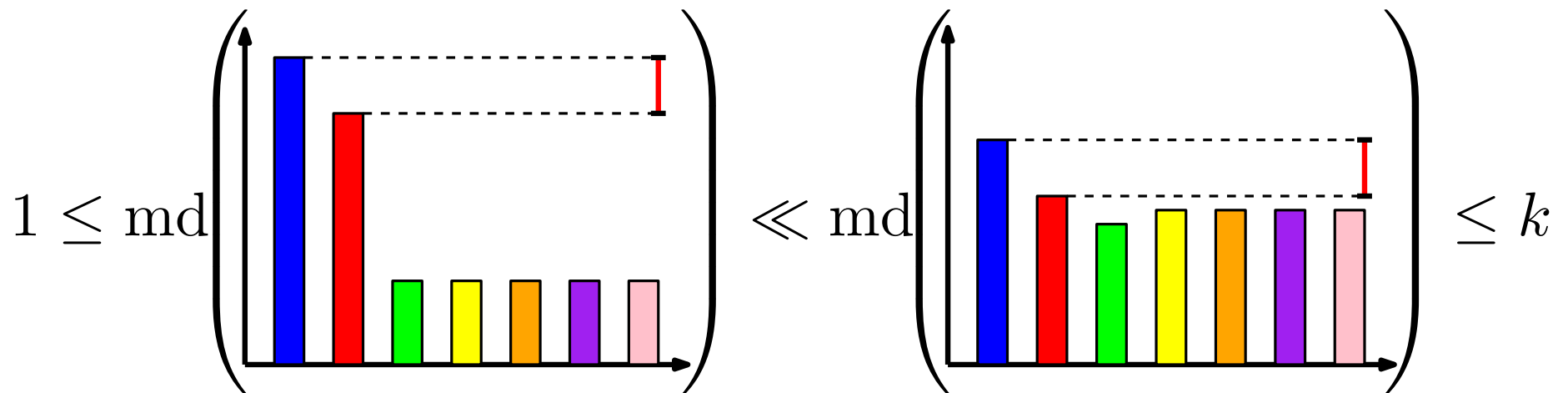
$c_i^{(t)} := \# \text{ nodes with color } i, \quad \mathbf{c}^{(t)} := \text{configuration at time } t.$

$$\text{md}(\mathbf{c}^{(0)}) := \sum_{i=1}^k \left( \frac{c_i^{(0)}}{c_1^{(0)}} \right)^2 = 1 + \mathcal{D} \left( \begin{array}{c} \text{Bar chart with 5 bars of different heights and colors (blue, red, yellow, green, orange)} \\ \text{Bar chart with 1 bar of height 1 (blue)} \end{array} \right)$$

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$$1 \leq \text{md} \left( \text{bar chart with 7 bars of varying heights} \right) \ll \text{md} \left( \text{bar chart with 7 bars of varying heights, all close to the maximum} \right) \leq k$$


# Convergence of the Undecided-State [SODA '15]

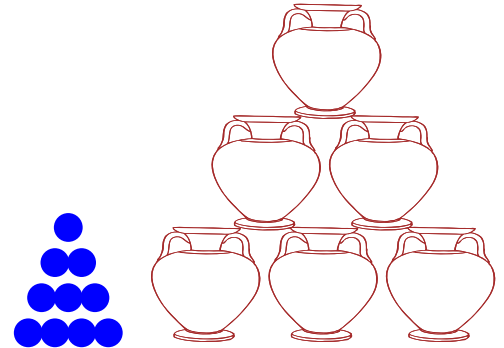
## Theorem

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

Thank You!

# Other Stuff with My Group

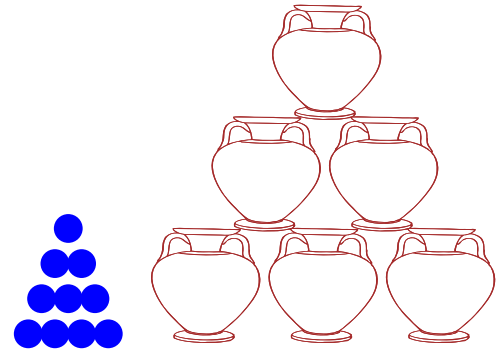
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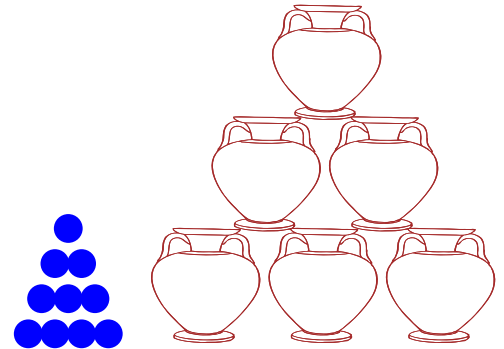
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Valid Almost-Consensus against  
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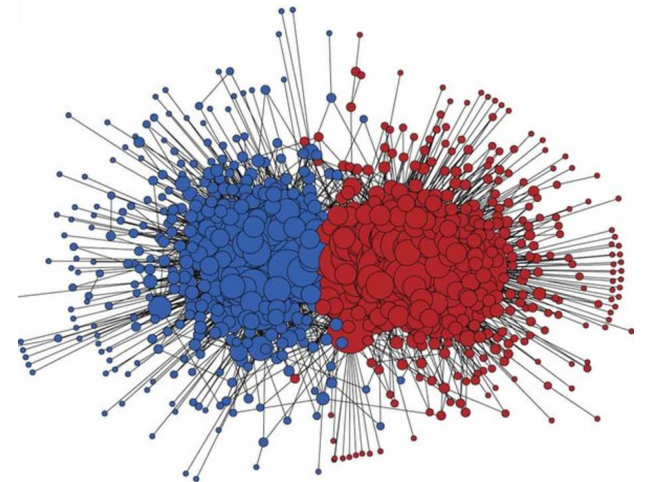
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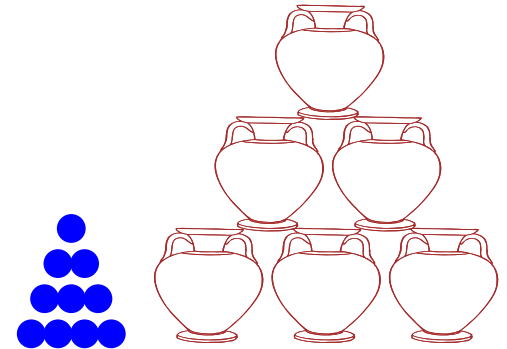
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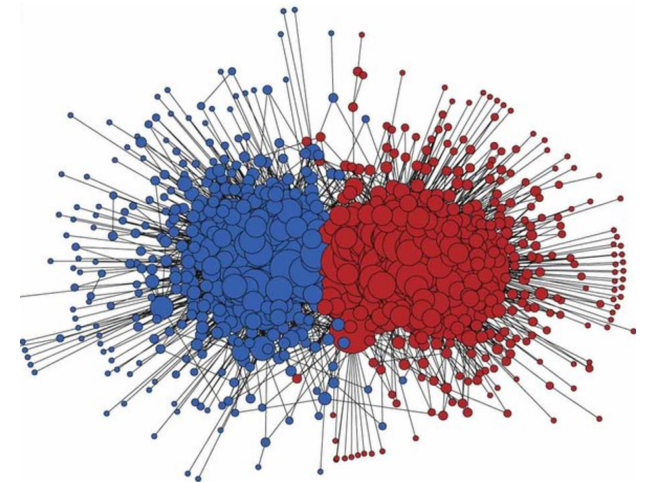
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Jan-May at the Simons Institute



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With L. Gualà and S. Leucci:  
NP-Hardness of Match-3 Games  
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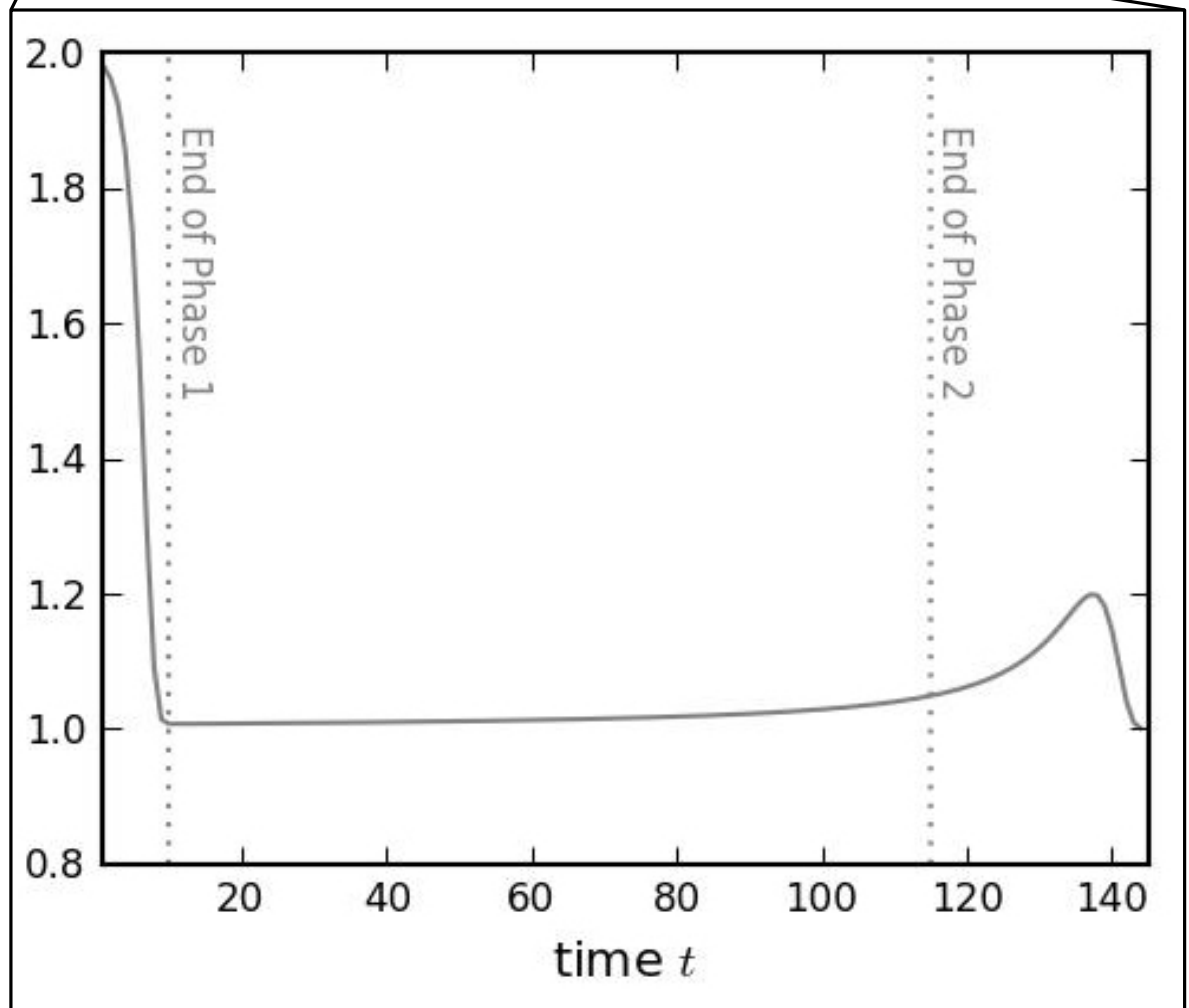
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With P. Fraigniaud: “Natural”  
Consensus with Noisy  
Communication [Coming soon]



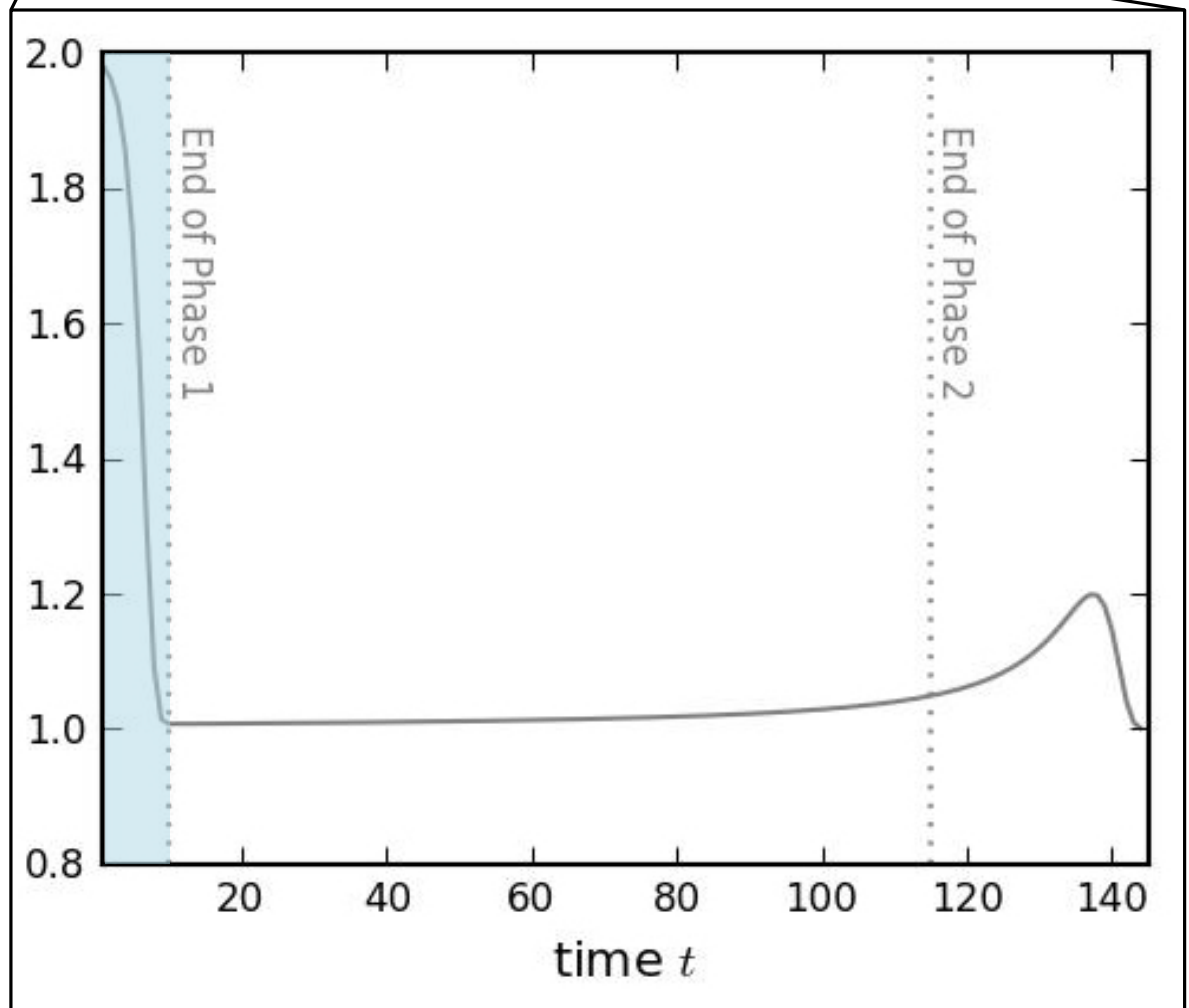
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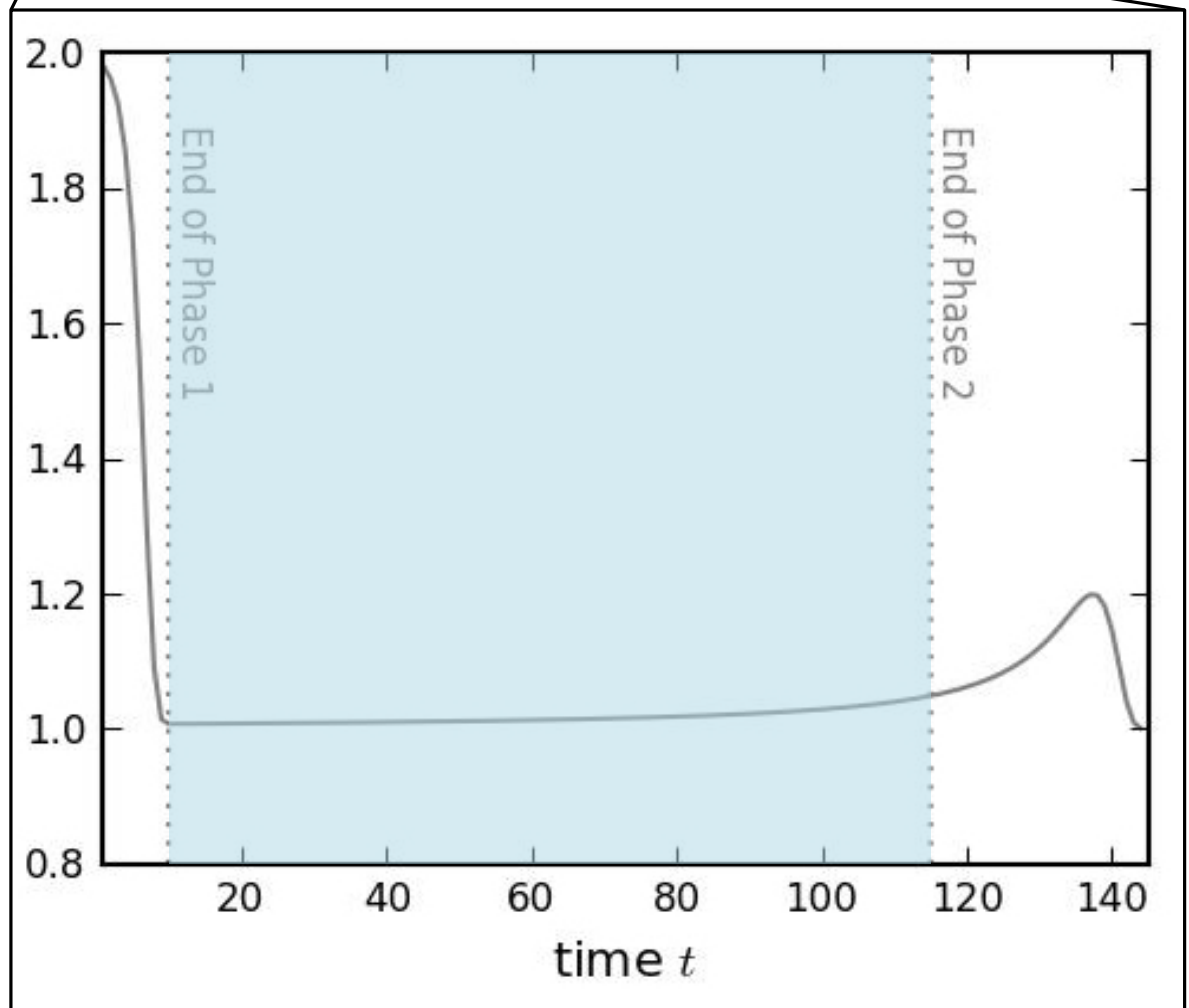
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