

# Molecular Computation

## An Algorithmic Approach

Rati Gelashvili

Joint work with

Dan Alistarh (ETH), David Eisenstat (Google),  
James Aspnes (Yale), Milan Vojnovic (MSR), Ron Rivest (MIT)

# Distributed Systems

Ingredients:



# Distributed Systems



Ingredients:

Nodes

-

# Distributed Systems



Ingredients:

Nodes

Communication

- 
-

# Distributed Systems



Ingredients:

- Nodes
- Communication
- Computation

# Computational Model

## Population Protocols [AADFP'04]



# Computational Model

## Population Protocols [AADFP'04]

- **Nodes** are *simple, identical agents*
  - Each node is *the same* finite state automaton
  - For example: a molecule



# Computational Model

## Population Protocols [AADFP'04]

- **Nodes** are *simple, identical agents*
  - Each node is *the same* finite state automaton
  - For example: a molecule
- **Interactions** are *pairwise*, and follow a *fair scheduler*
  - Usually considered *uniform random*
  - Nodes update their state following interactions

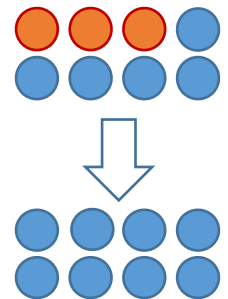




# Computational Model

## Population Protocols [AADFP'04]

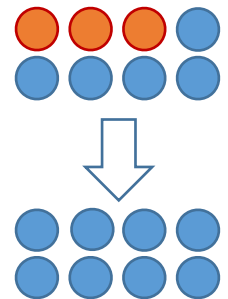
- **Nodes** are *simple, identical agents*
  - Each node is *the same* finite state automaton
  - For example: a molecule
- **Interactions** are *pairwise*, and follow a *fair scheduler*
  - Usually considered *uniform random*
  - Nodes update their state following interactions
- **Computation** is performed *collectively*
  - The system *should converge* to configurations satisfying meaningful predicates
  - No “fixed” decision time



# Computational Model

## Population Protocols [AADFP'04]

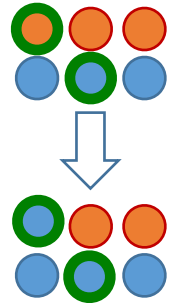
- **Nodes** are *simple, identical agents*
  - Each node is *the same* finite state automaton
  - For example: a molecule
- **Interactions** are *pairwise*, and follow a *fair scheduler*
  - Usually considered *uniform random*
  - Nodes update their state following interactions
- **Computation** is performed *collectively*
  - The system *should converge* to configurations satisfying meaningful predicates
  - No “fixed” decision time
- A.k.a. Chemical Reaction Networks



# Complexity

## 1. Time

- **Round** = a *single pair* interacts
  - Chosen uniformly at random
- **Parallel convergence time**
  - #rounds to convergence / # nodes
  - Alternative continuous-time definition exists



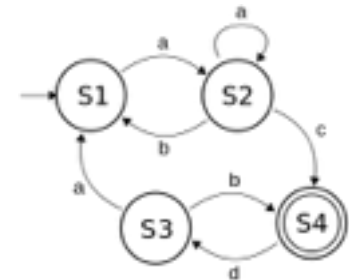
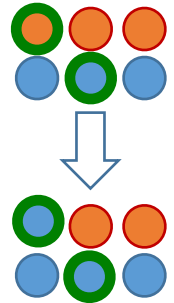
# Complexity

## 1. Time

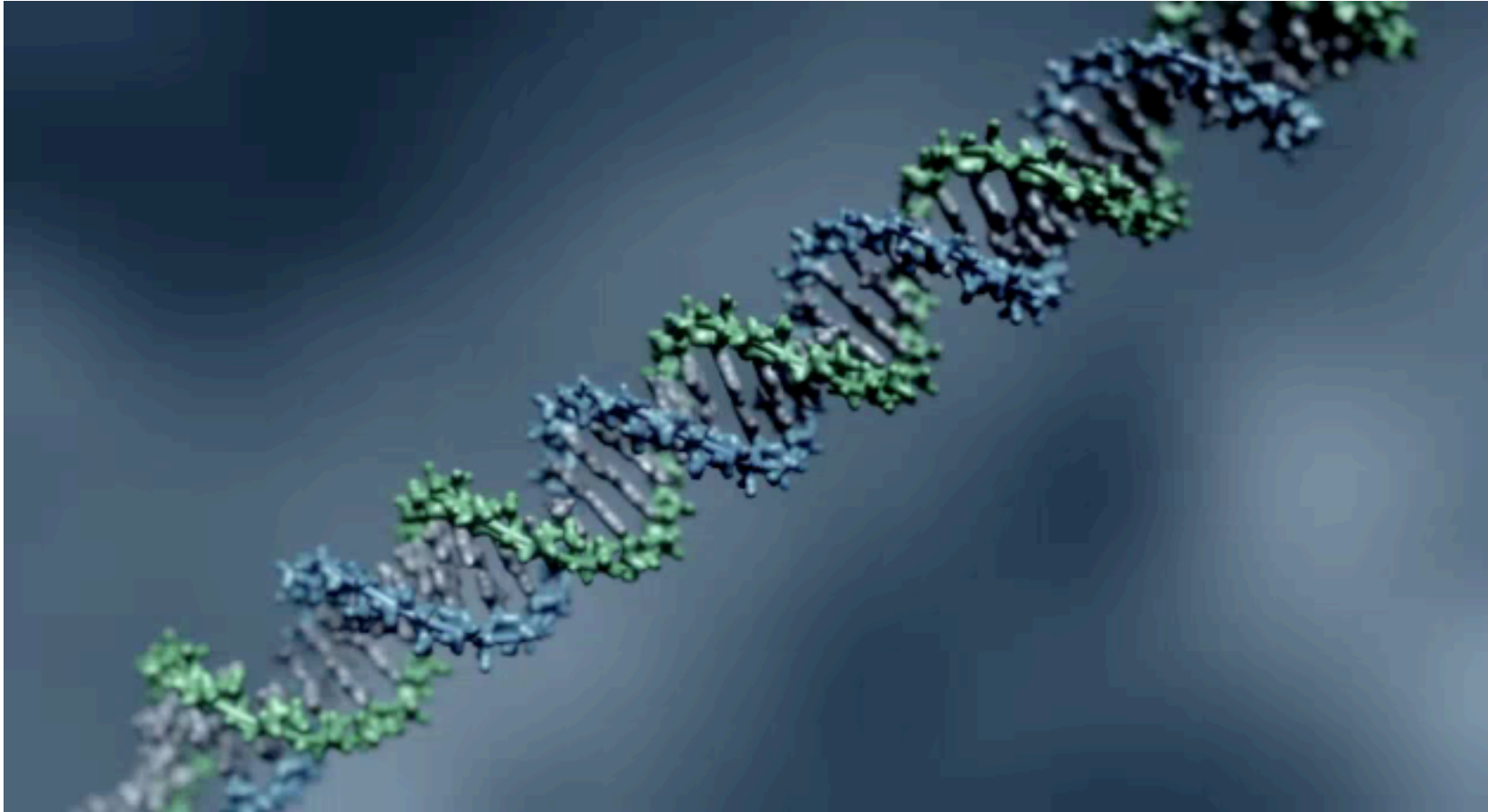
- **Round** = a *single pair* interacts
  - Chosen uniformly at random
- **Parallel convergence time**
  - #rounds to convergence / # nodes
  - Alternative continuous-time definition exists

## 2. Space

- Number of *distinct states* per automaton
- Alternatively, #memory bits to encode state

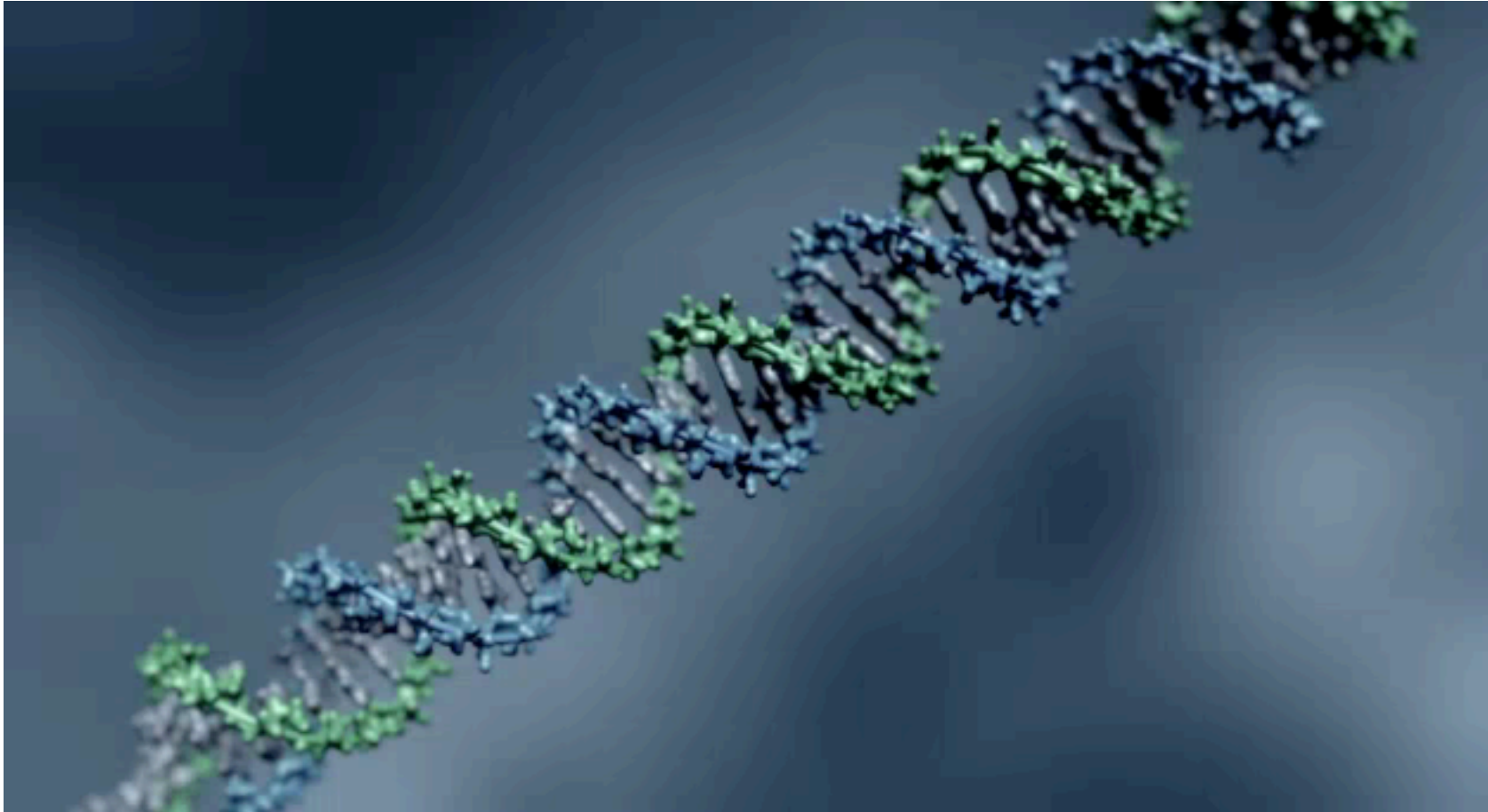


# More Precisely: Communication



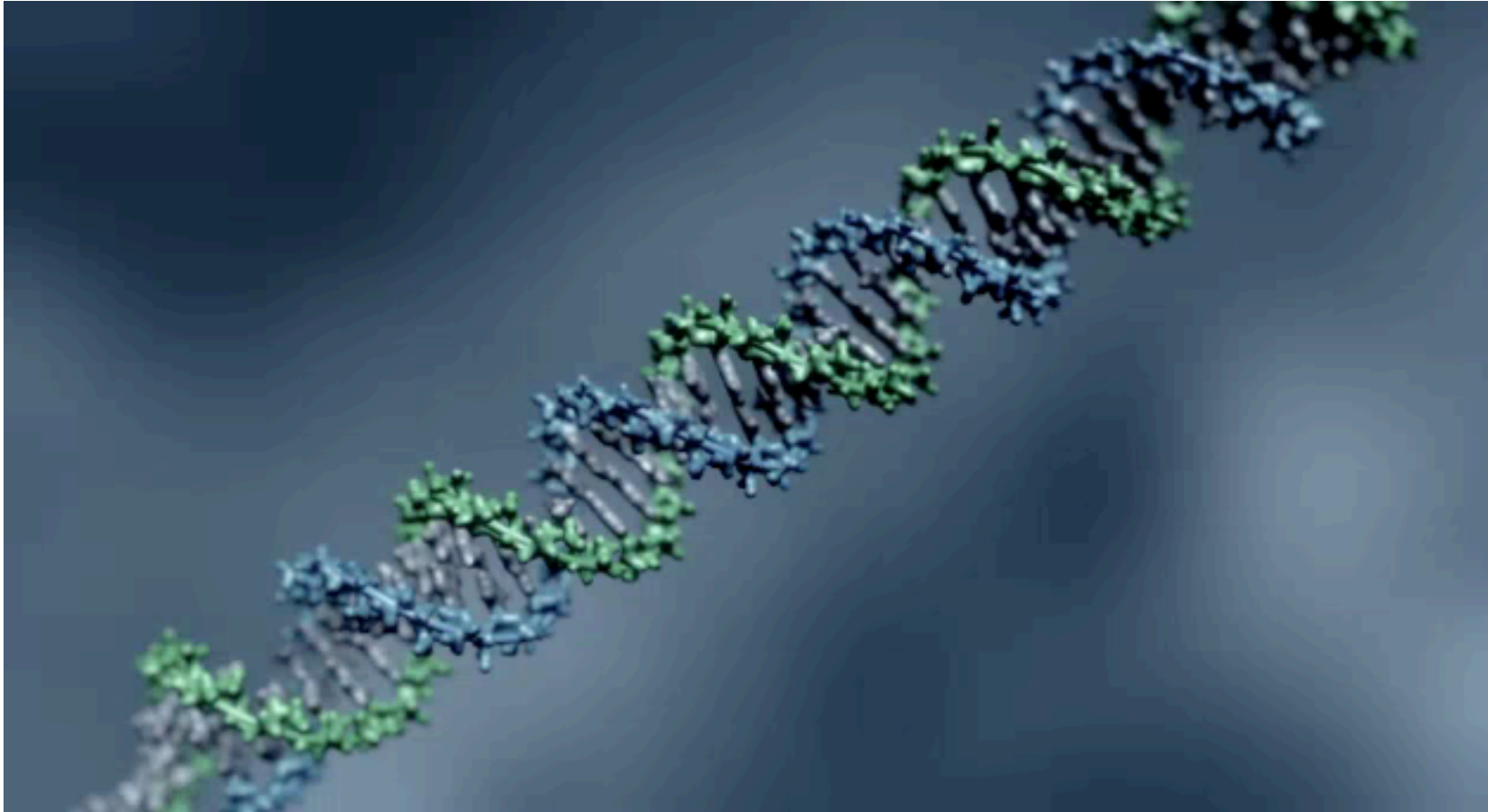
Courtesy of the Microsoft Research Biological Computation Group

# More Precisely: Communication



Courtesy of the Microsoft Research Biological Computation Group

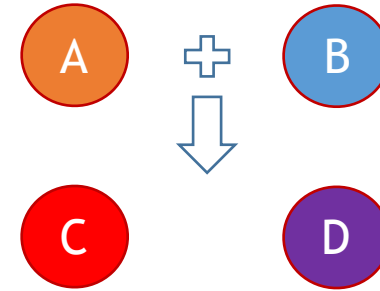
# More Precisely: Communication



Courtesy of the Microsoft Research Biological Computation Group

# What can we compute?

We can perform interactions of the type:



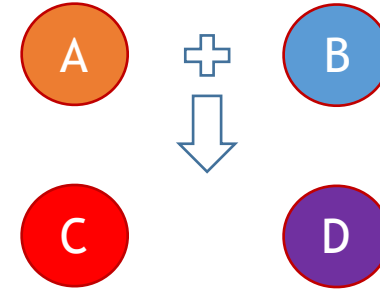


# What can we compute?

We can perform interactions of the type:

Example: the **OR** function

- **Initial states:** 0 or 1
- **Final state:**
  - If there exists a 1, then all 1.
  - Otherwise, all 0
- **Protocol:**

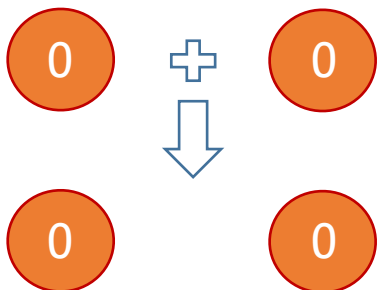
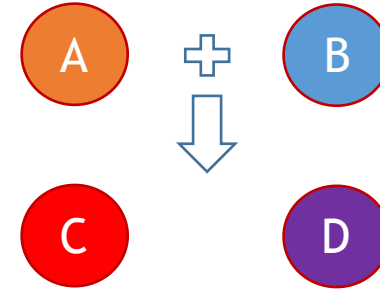


# What can we compute?

We can perform interactions of the type:

Example: the **OR** function

- **Initial states:** 0 or 1
- **Final state:**
  - If there exists a 1, then all 1.
  - Otherwise, all 0
- **Protocol:**

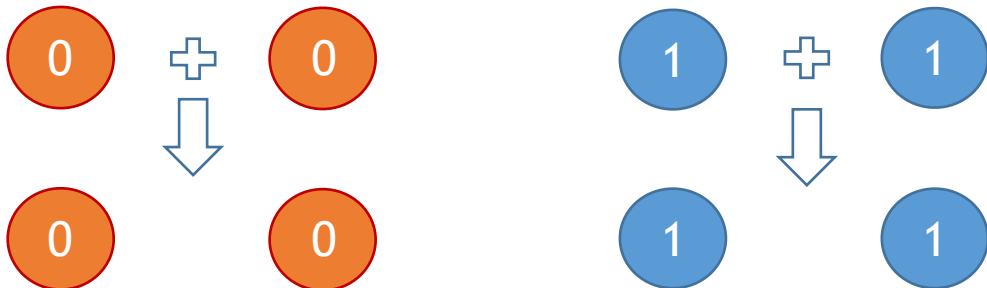
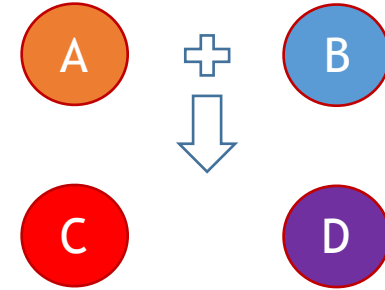


# What can we compute?

We can perform interactions of the type:

Example: the **OR** function

- **Initial states:** 0 or 1
- **Final state:**
  - If there exists a 1, then all 1.
  - Otherwise, all 0
- **Protocol:**

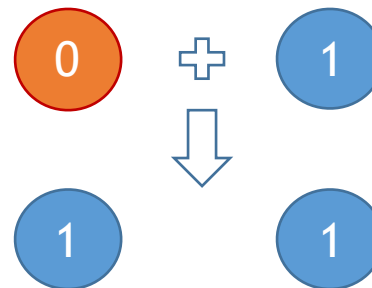
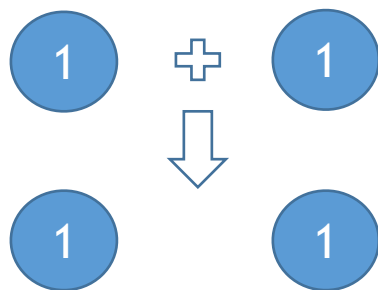
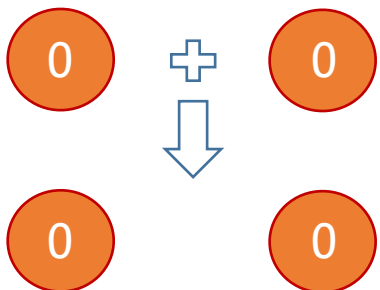
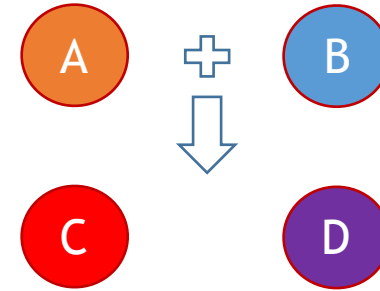


# What can we compute?

We can perform interactions of the type:

Example: the **OR** function

- **Initial states:** 0 or 1
- **Final state:**
  - If there exists a 1, then all 1.
  - Otherwise, all 0
- **Protocol:**

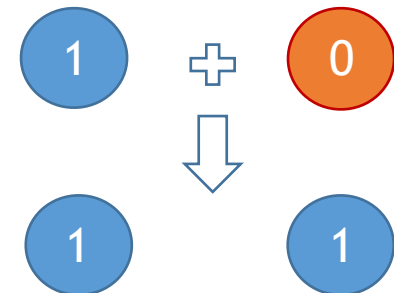
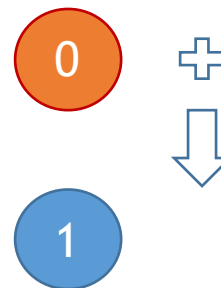
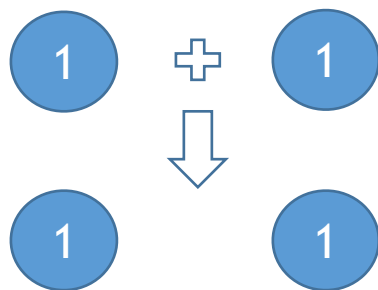
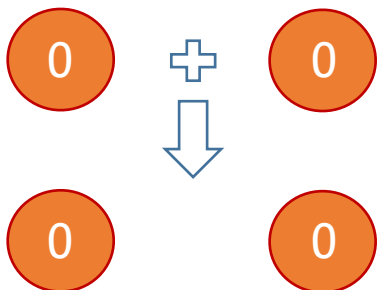
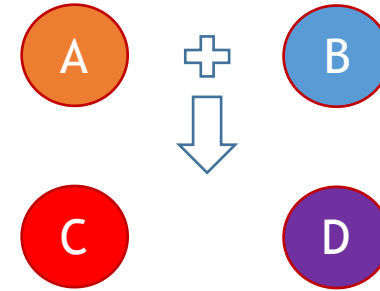


# What can we compute?

We can perform interactions of the type:

Example: the **OR** function

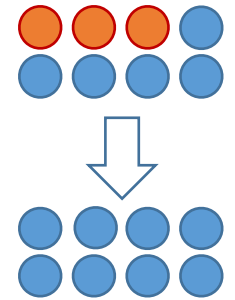
- **Initial states:** 0 or 1
- **Final state:**
  - If there exists a 1, then all 1.
  - Otherwise, all 0
- **Protocol:**



# The Majority Function

Majority (“Consensus”)

- Initial states A, B
- Output:
  - A if  $\#A > \#B$  initially.
  - B, otherwise.

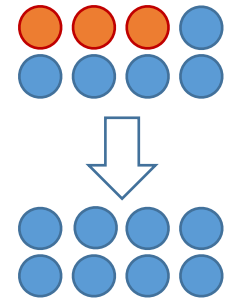


# The Majority Function



## Majority (“Consensus”)

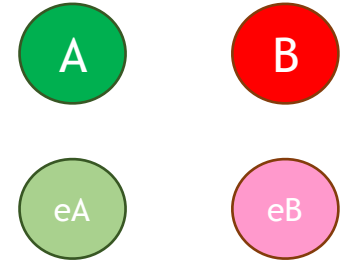
- **Initial states A, B**
- **Output:**
  - A if  $\#A > \#B$  initially.
  - B, otherwise.
- **Fundamental task**
  - **Complexity:** [AAE08] & [DV12]; [PVV09] & [MNRS14]
  - Natural computation:  
**the cell cycle switch implements approximate majority [CC12]**
  - Implementation in DNA: [CDS<sup>+</sup>13, Nature Nanotechnology]



# Solving Majority

4-State Exact Majority [PVV09] [MNRS14]

- Protocol:

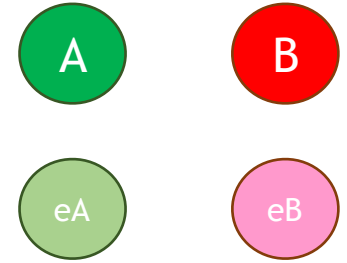
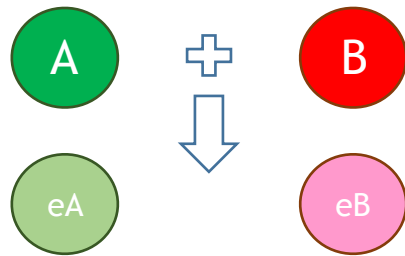




# Solving Majority

4-State Exact Majority [PVV09] [MNRS14]

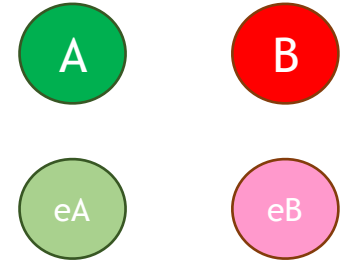
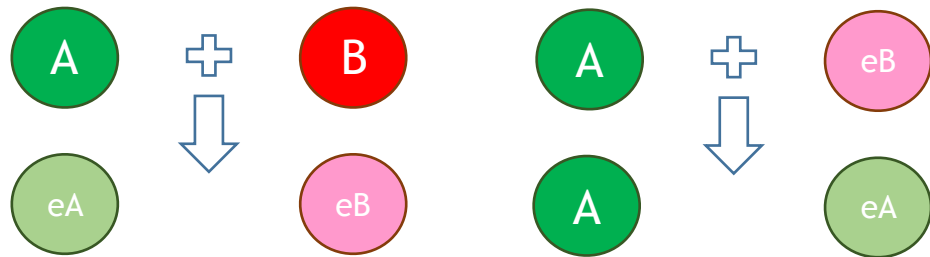
- Protocol:



# Solving Majority

## 4-State Exact Majority [PVV09] [MNRS14]

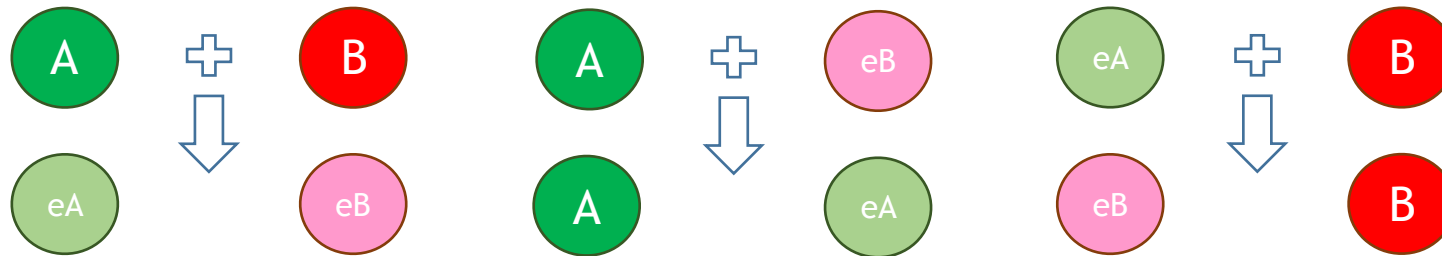
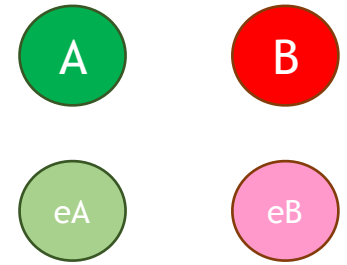
- Protocol:



# Solving Majority

## 4-State Exact Majority [PVV09] [MNRS14]

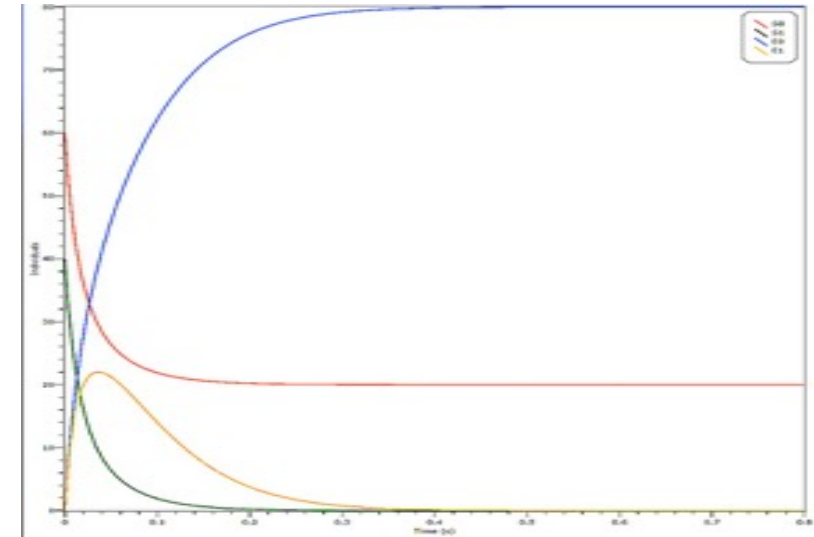
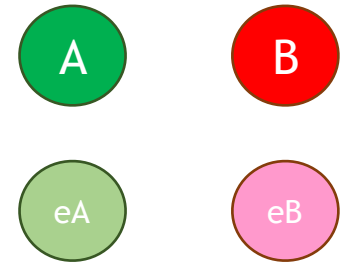
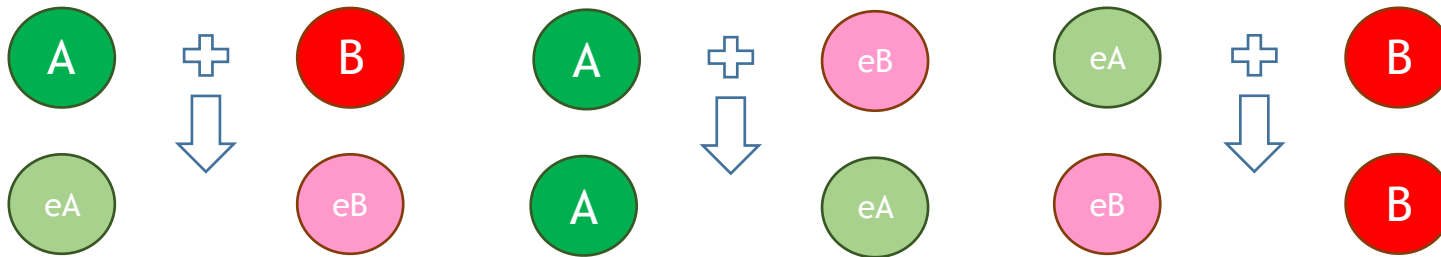
- Protocol:



# Solving Majority

## 4-State Exact Majority [PVV09] [MNRS14]

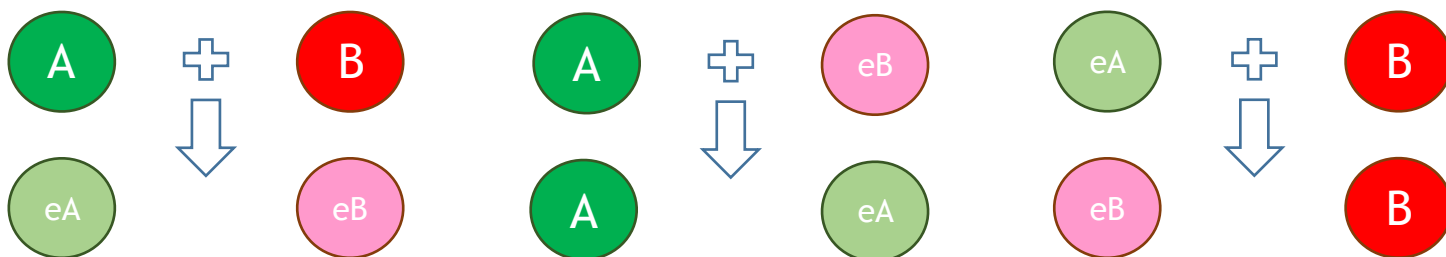
- Protocol:



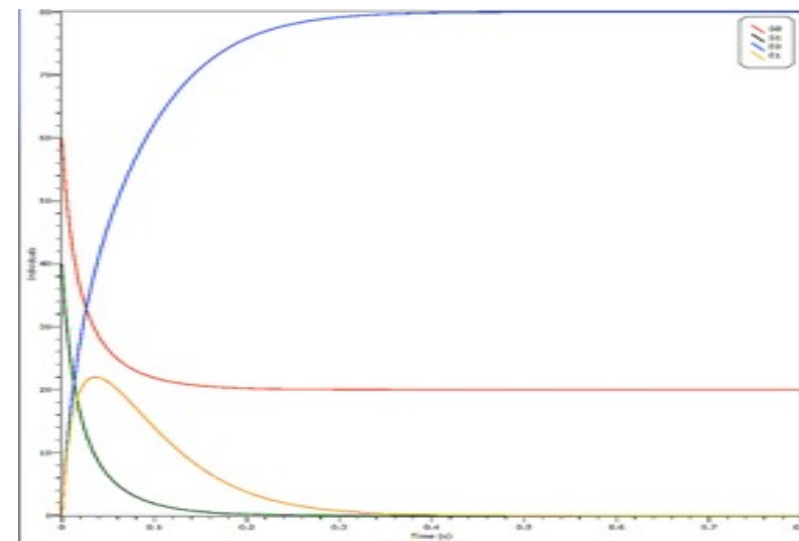
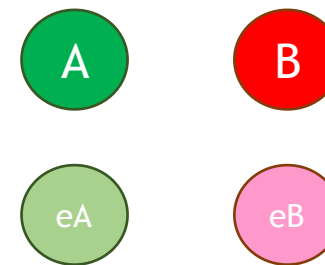
# Solving Majority

## 4-State Exact Majority [PVV09] [MNRS14]

- Protocol:



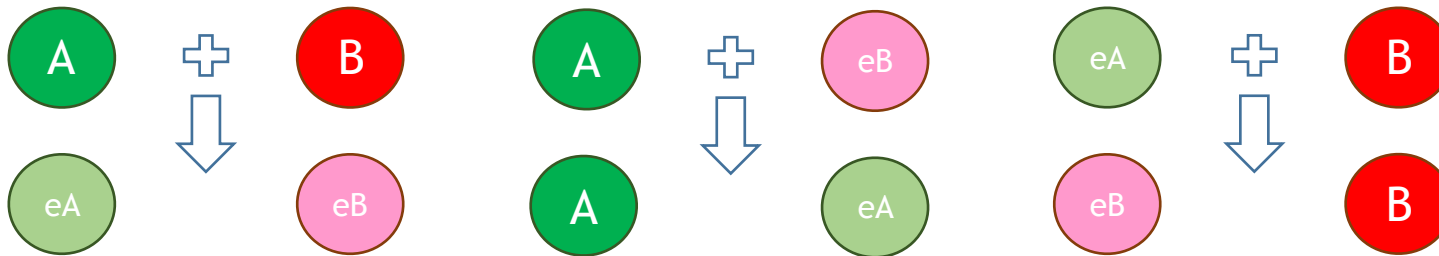
Discrepancy/margin:  
 $\epsilon = |\#A - \#B| / n$   
Can be as small as  
 $\epsilon = O(1 / n)$ .



# Solving Majority

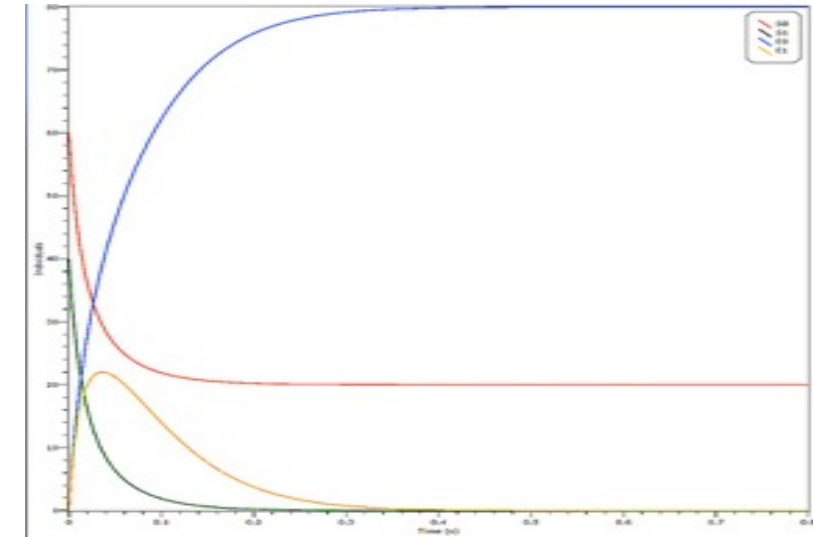
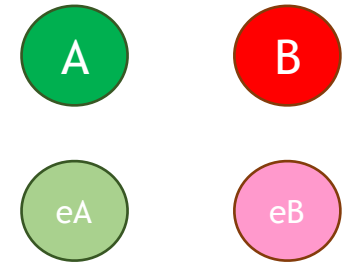
## 4-State Exact Majority [PVV09] [MNRS14]

- Protocol:



Discrepancy/margin:  
 $\epsilon = |\#A - \#B| / n$   
 Can be as small as  
 $\epsilon = O(1 / n)$ .

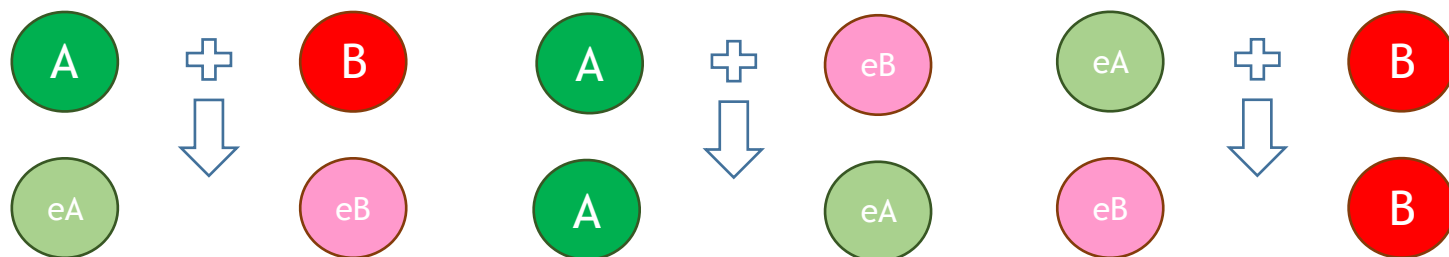
Theorem: Given  $n$  nodes and discrepancy  $\epsilon$ , the running time of 4EM is  $O((\log n) / \epsilon)$ .



# Solving Majority

## 4-State Exact Majority [PVV09] [MNRS14]

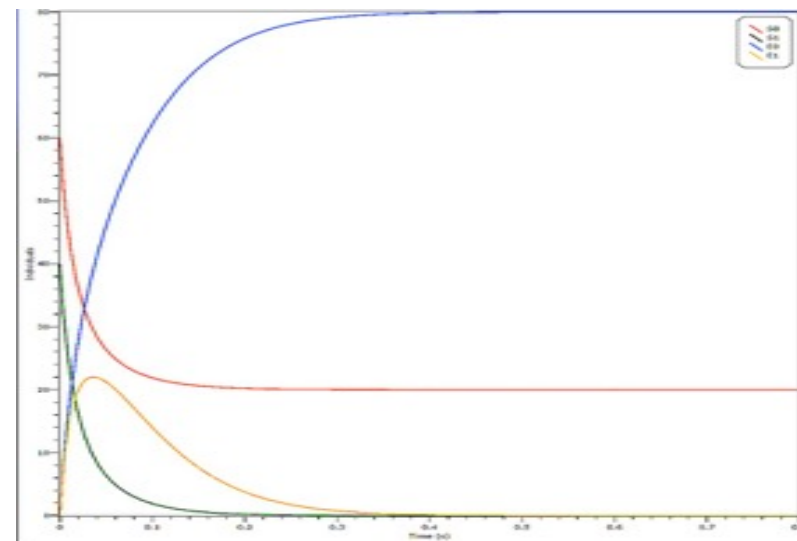
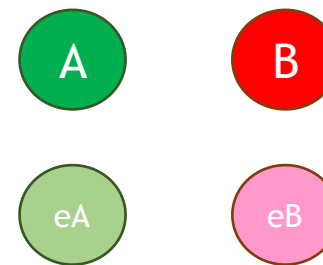
- Protocol:



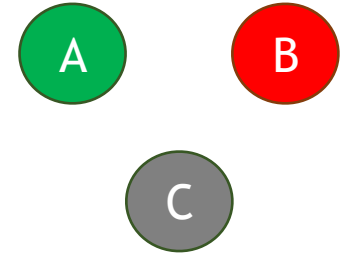
Discrepancy/margin:  
 $\epsilon = |\#A - \#B| / n$   
Can be as small as  
 $\epsilon = O(1 / n)$ .

Theorem: Given  $n$  nodes and discrepancy  $\epsilon$ , the running time of 4EM is  $O((\log n) / \epsilon)$ .

Can be  $\Theta(n \log n)$  if  $\epsilon = \text{constant} / n$ .



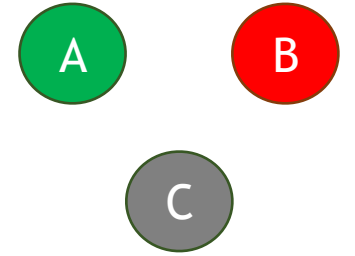
# Solving Majority *Approximately*



- 3-state Approximate Majority [AAE08] [DV12]

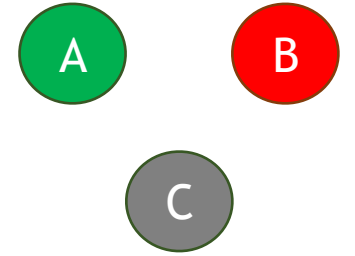


# Solving Majority *Approximately*

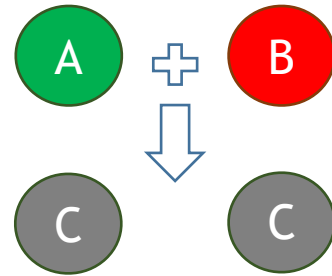


- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:

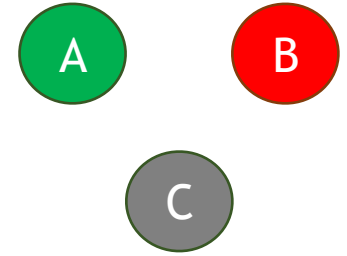
# Solving Majority *Approximately*



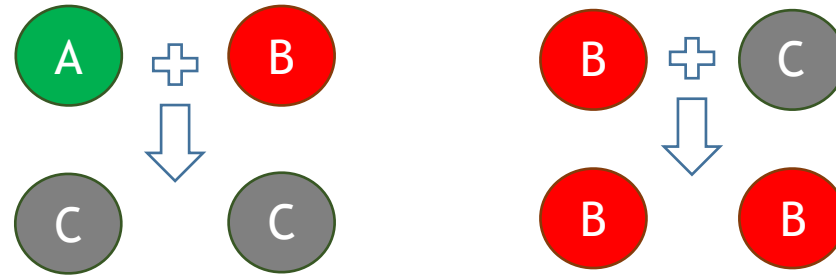
- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:



# Solving Majority *Approximately*

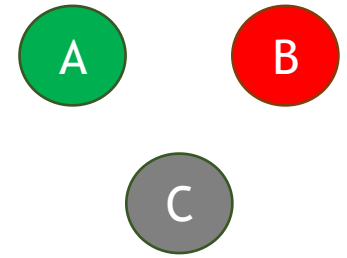
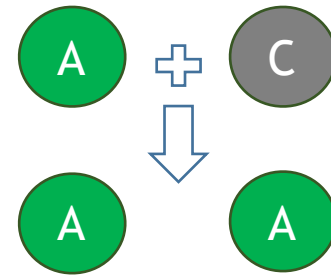
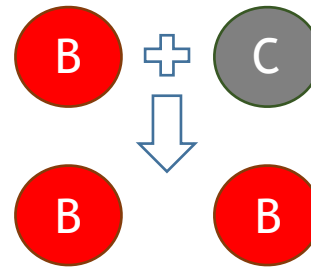
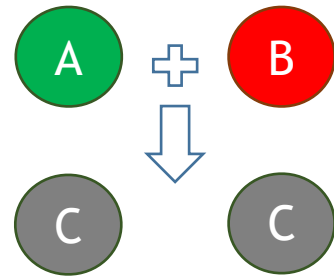


- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:

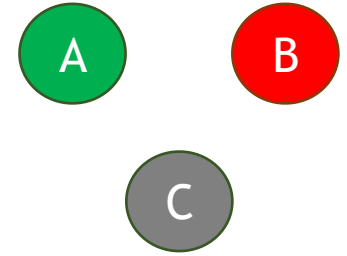


# Solving Majority *Approximately*

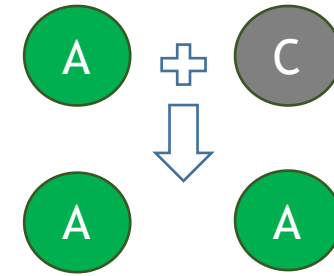
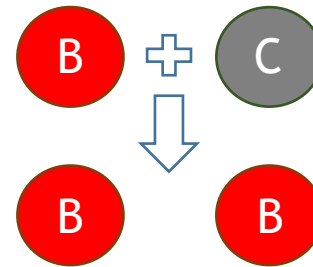
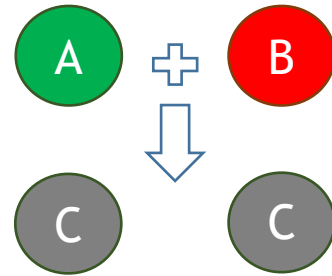
- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:



# Solving Majority *Approximately*

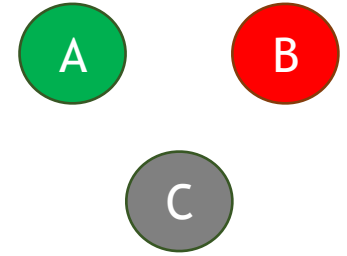


- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:

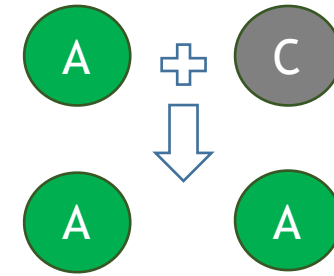
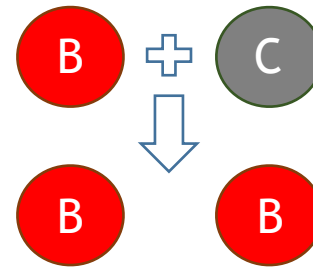
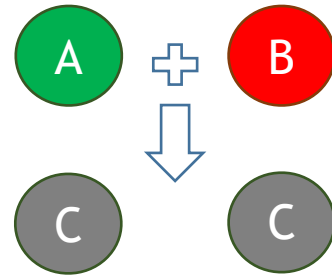


- Execution:

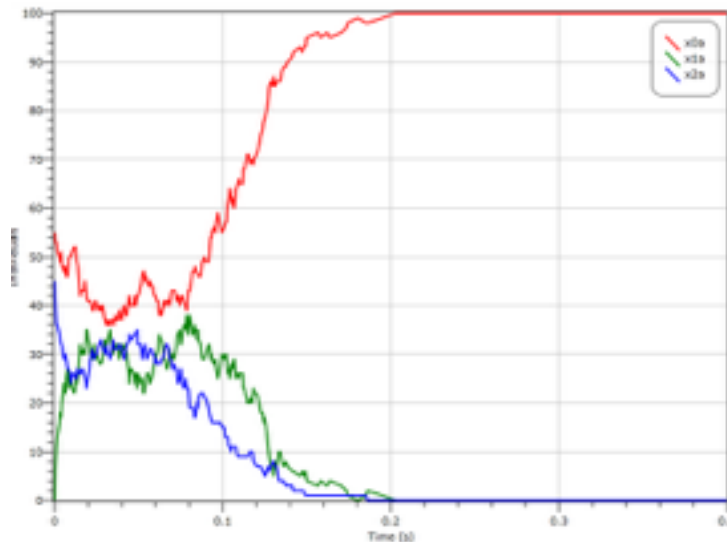
# Solving Majority *Approximately*



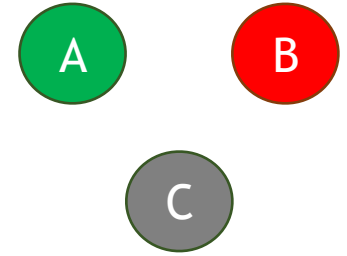
- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:



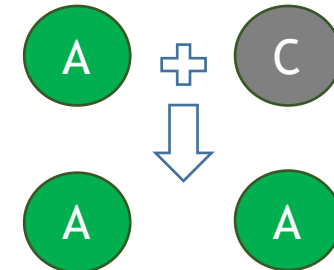
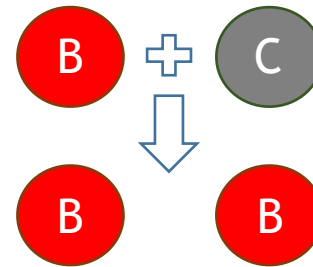
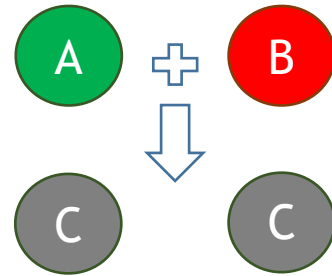
- Execution:



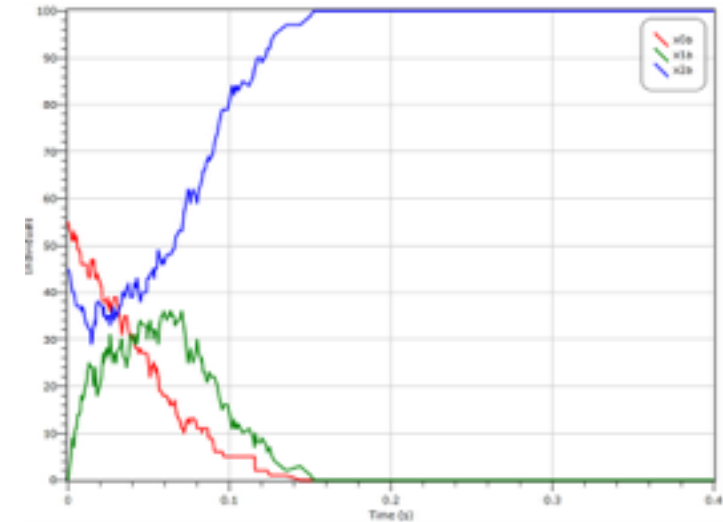
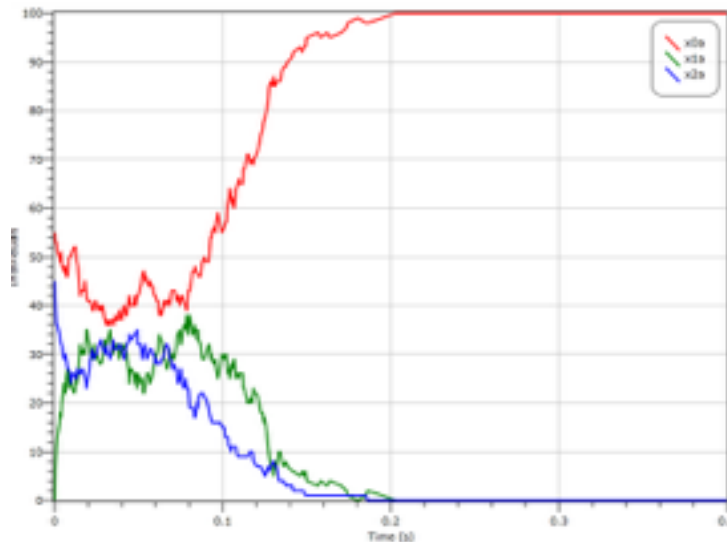
# Solving Majority *Approximately*



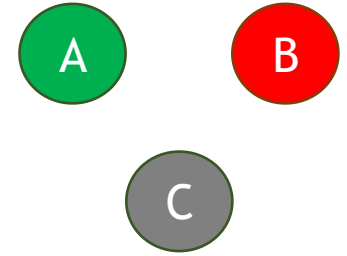
- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:



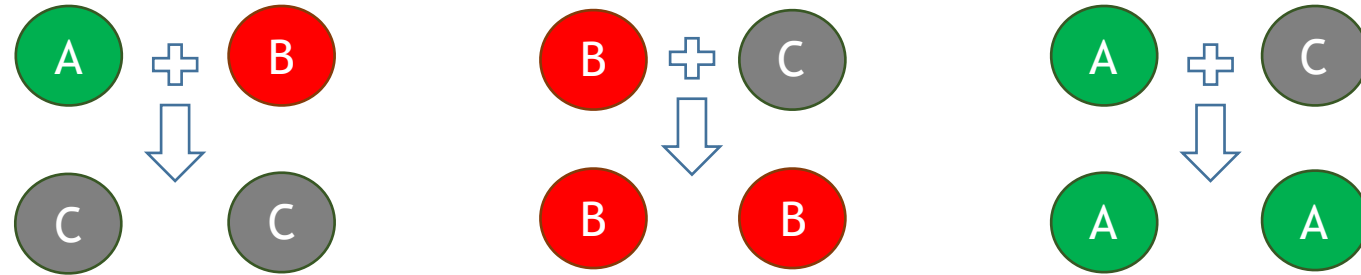
- Execution:



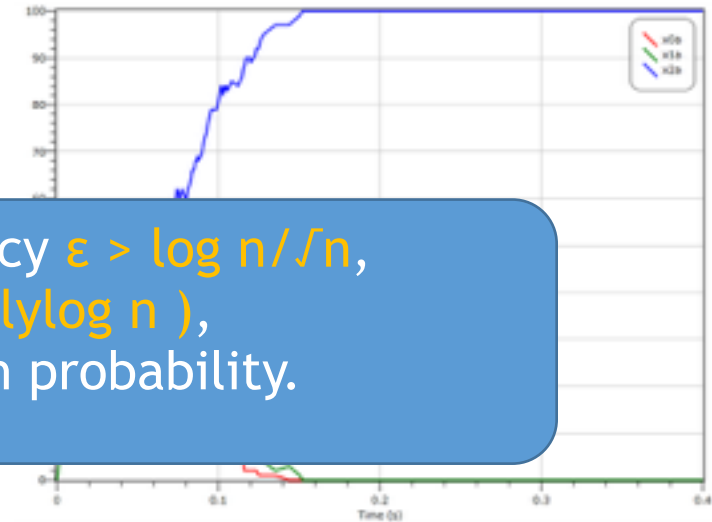
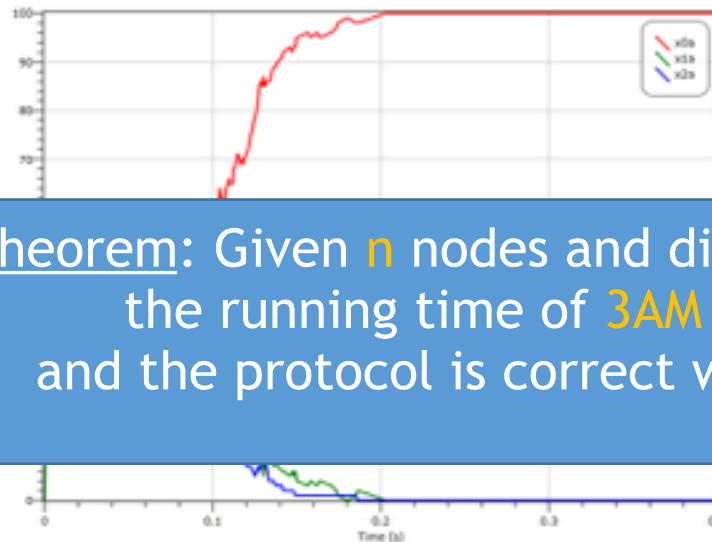
# Solving Majority *Approximately*



- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:



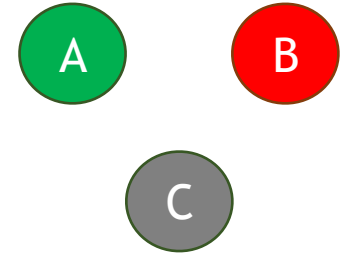
- Execution:



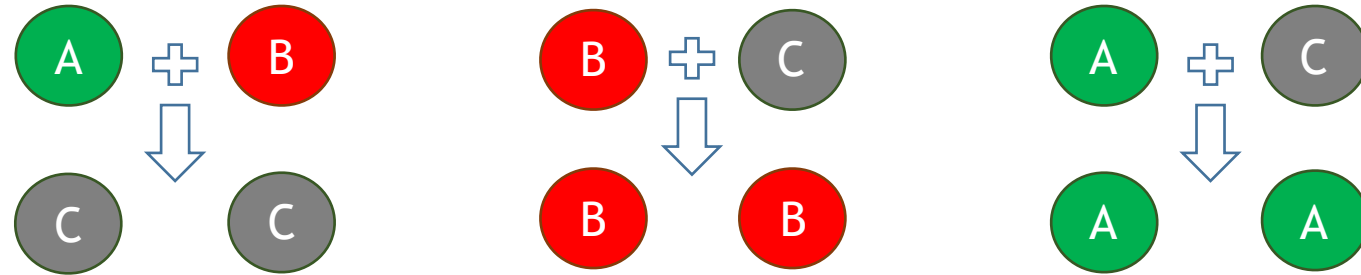
Theorem: Given  $n$  nodes and discrepancy  $\epsilon > \log n / \sqrt{n}$ , the running time of 3AM is  $O(\text{polylog } n)$ , and the protocol is correct with high probability.



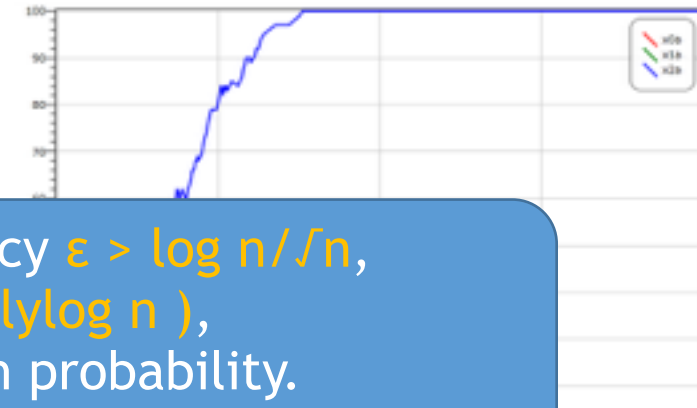
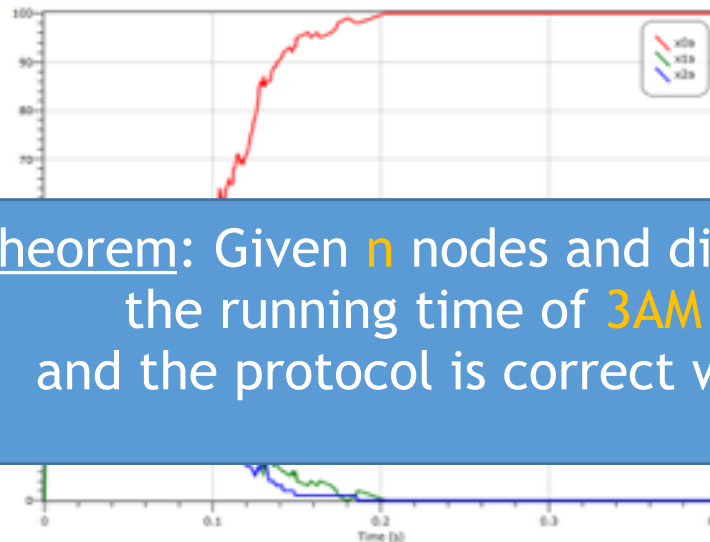
# Solving Majority *Approximately*



- 3-state Approximate Majority [AAE08] [DV12]
- The protocol:



- Execution:



Theorem: Given  $n$  nodes and discrepancy  $\epsilon > \log n / \sqrt{n}$ , the running time of 3AM is  $O(\text{polylog } n)$ , and the protocol is correct with high probability.

Error probability can be as high as *constant* for lower discrepancy.

# The Status

Algorithm	Reliability	Speed
The Four-State Protocol	Exact	Slow (super-linear)
The Three-State Protocol	Flaky (Up to Constant Error)	Fast (poly-logarithmic)

# Average&Conquer

Algorithm	Reliability	Speed
The Four-State Protocol	Exact	Slow (super-linear)
The Three-State Protocol	Flaky (Up to Constant Error)	Fast (poly-logarithmic)
Average&Conquer [PODC 2015]	Exact	Fast (poly-logarithmic)

# Average&Conquer

Algorithm	Reliability	Speed
The Four-State Protocol	Exact	Slow (super-linear)
The Three-State Protocol	Flaky (Up to Constant Error)	Fast (poly-logarithmic)
Average&Conquer [PODC 2015] (Super-Constant State Space)	Exact	Fast (poly-logarithmic)

# The Plan

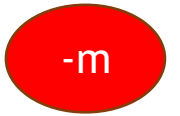
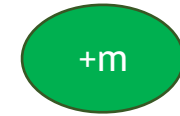
- Population Protocols
- The Majority Problem
  - 4EM
  - 3AM
  - Average-and-Conquer (AVC)
  - Quantized AVC
- Impossibility Results
- Open Questions
- Leader Election Problem



# Simplified AVC: Main Ideas



- Each **state** corresponds to a **value** (“confidence level”)
  - **Strong states** (non-negative value):
    - Positive  $\rightarrow$  A
    - Negative  $\rightarrow$  B
  - **Weak:** value  $\pm 0$
- All nodes start with absolute value  $m > 0$ 
  - $+m$  if A
  - $-m$  if B
- Two interaction types:
  - **Averaging:** strong (non-zero) nodes average out their values
  - **Conquer:** strong (non-zero) nodes bring weak nodes to “their side”
- **Output:**
  - If **positive** or  $+0$ , then A
  - If **negative** or  $-0$ , then B

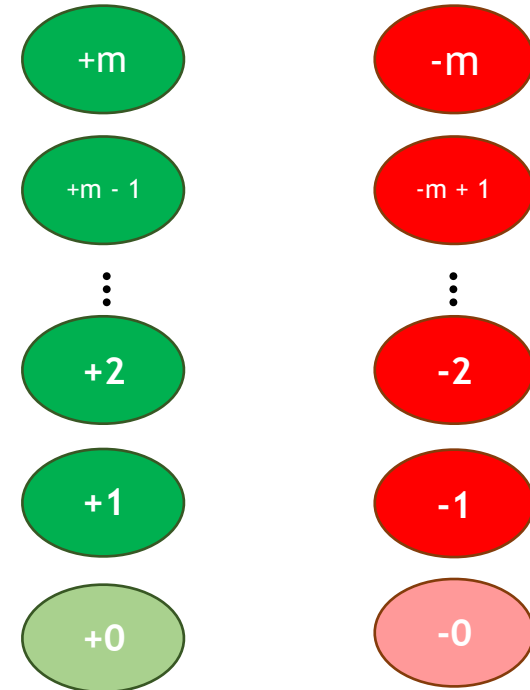


# AVC in Action

Initially:  $+m$  or  $-m$ , **odd** integers

**Strong** states: non-zero absolute value.

**Weak** states: value zero (+/-).



# AVC in Action

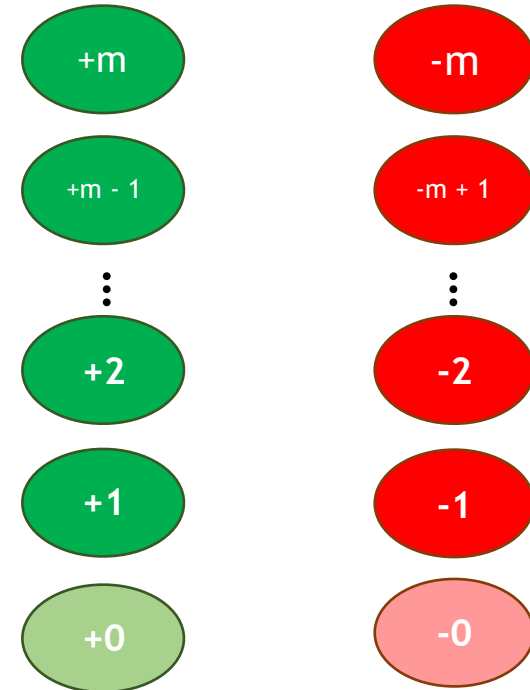
Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values





# AVC in Action

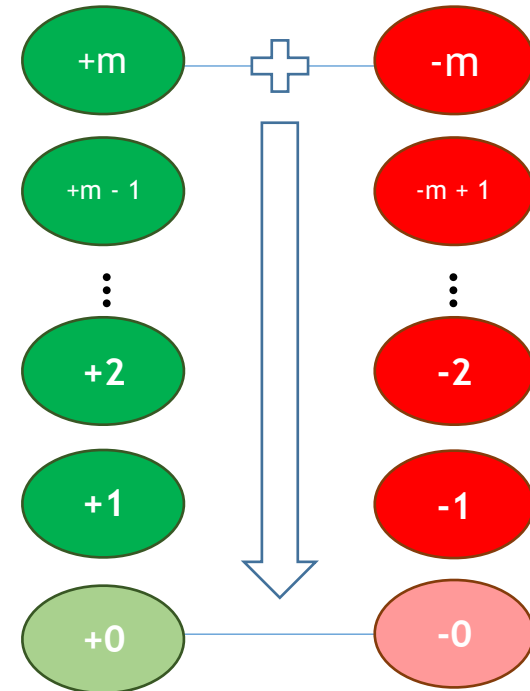
**Initially:**  $+m$  or  $-m$ , **odd** integers

**Strong states:** non-zero absolute value.

**Weak states:** value zero (+/-).

**Averaging:**

- Whenever two strong nodes meet, they average values



# AVC in Action

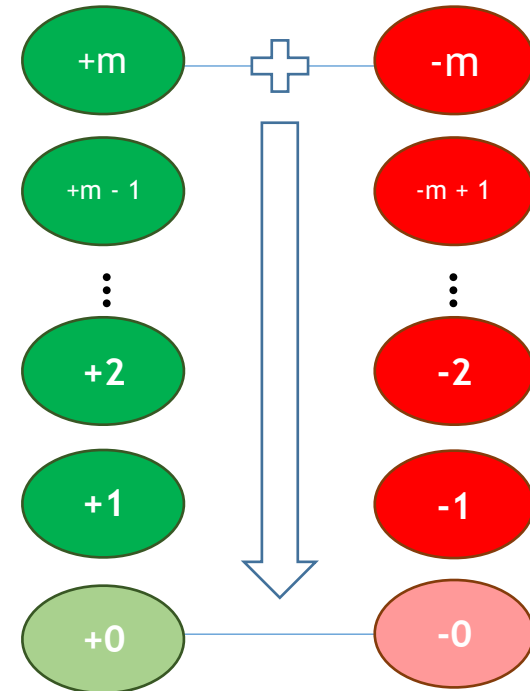
**Initially:**  $+m$  or  $-m$ , **odd** integers

**Strong states:** non-zero absolute value.

**Weak states:** value zero (+/-).

**Averaging:**

- Whenever two strong nodes meet, they average values



# AVC in Action

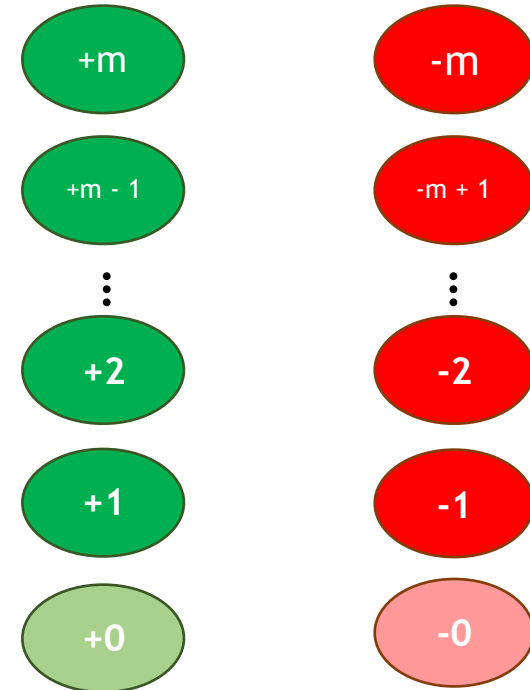
Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values



# AVC in Action

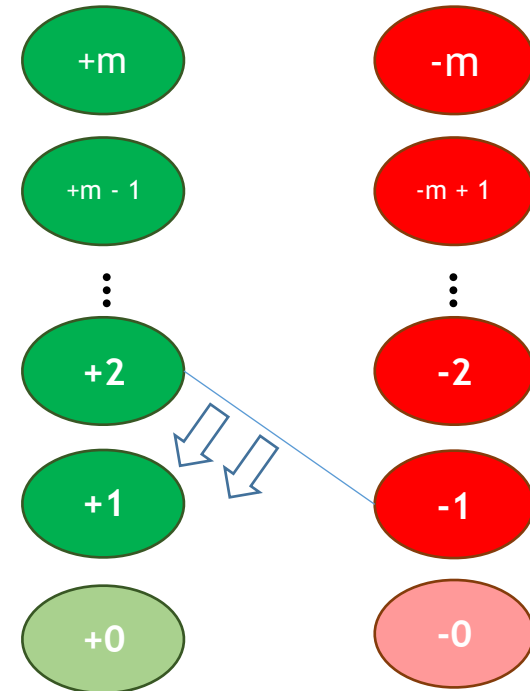
Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values



# AVC in Action

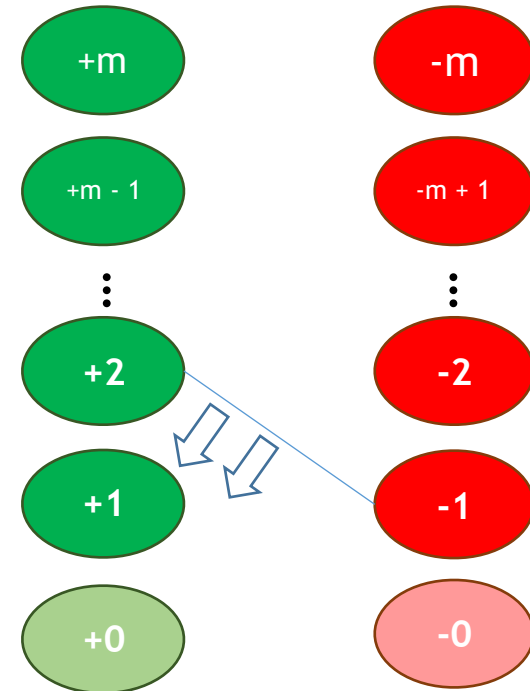
Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values



# AVC in Action

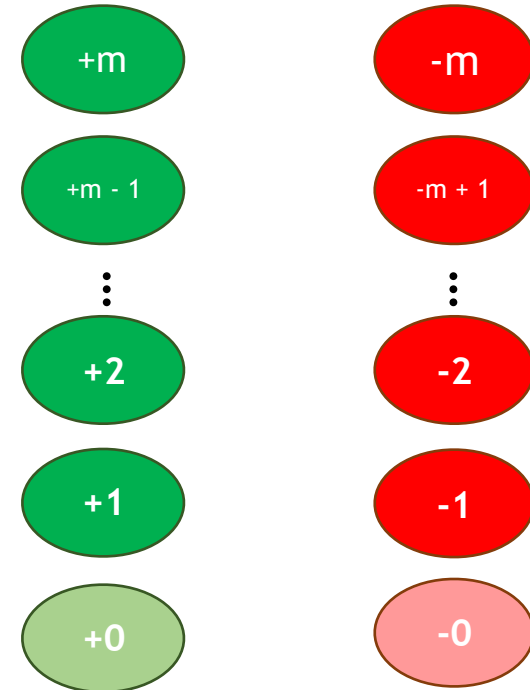
Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values



# AVC in Action

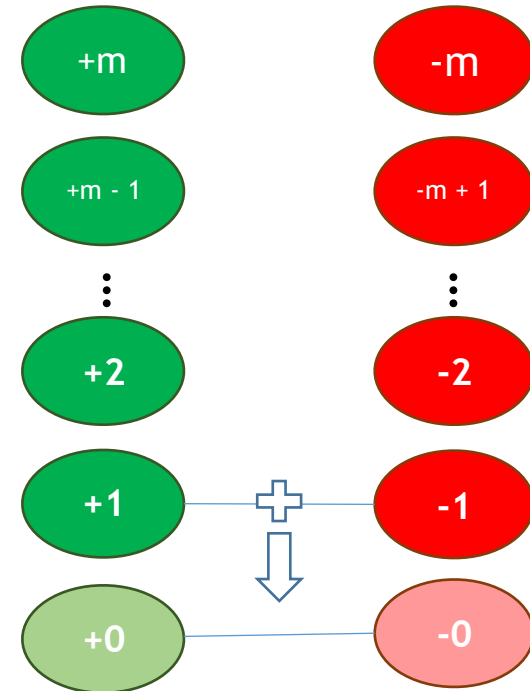
Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values



# AVC in Action

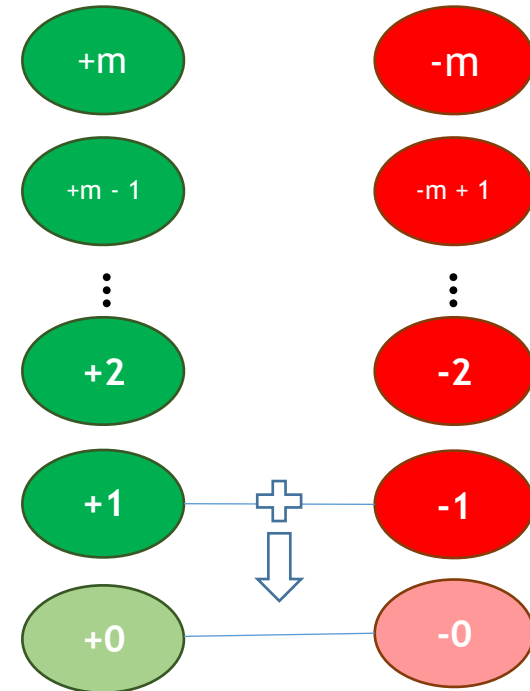
Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values





# AVC in Action

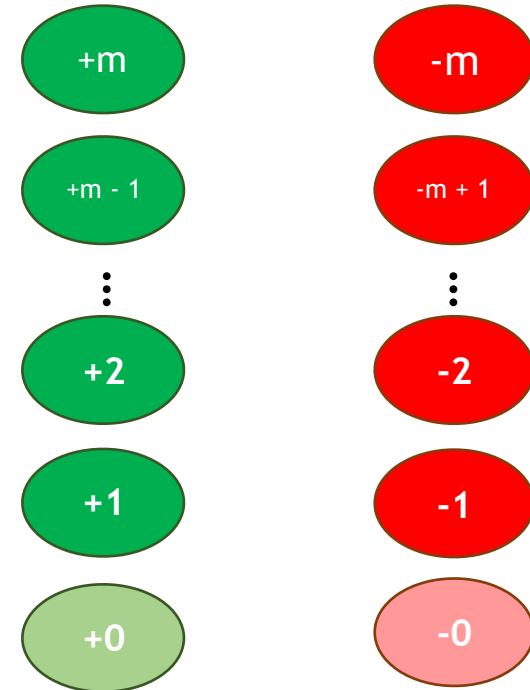
Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values



# AVC in Action

Initially:  $+m$  or  $-m$ , **odd** integers

**Strong** states: non-zero absolute value.

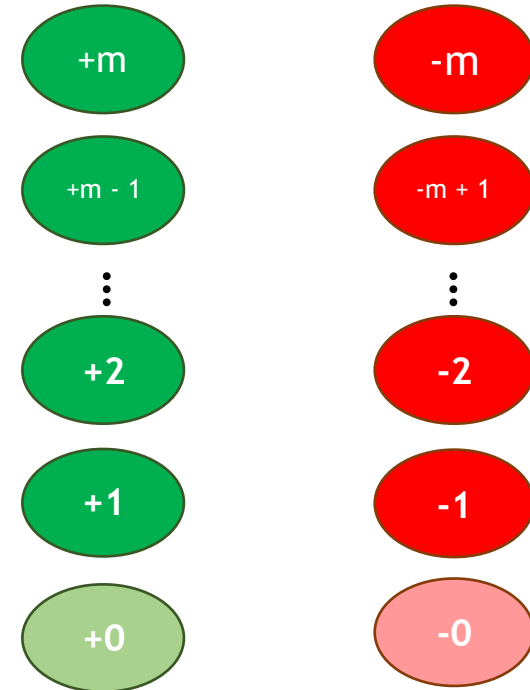
**Weak** states: value zero (+/-).

**Averaging:**

- Whenever two strong nodes meet, they average values

**Conquer:**

- **Strong** nodes sway **weak** nodes towards their decision.



# AVC in Action

Initially:  $+m$  or  $-m$ , **odd** integers

**Strong** states: non-zero absolute value.

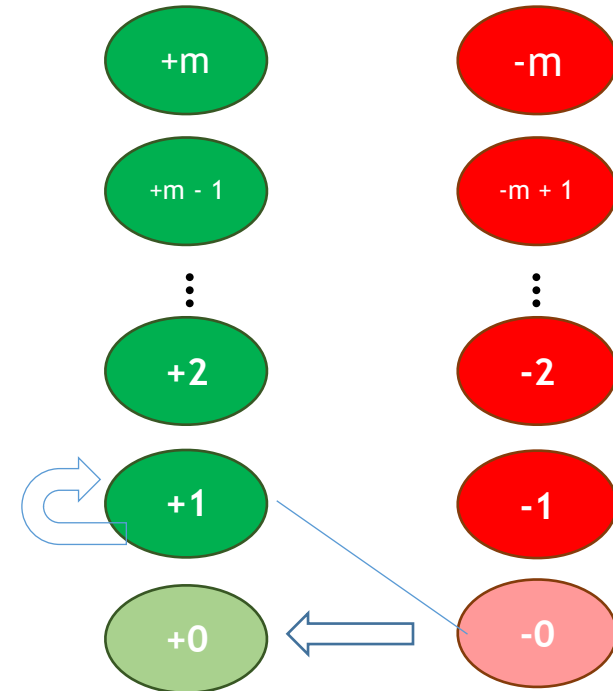
**Weak** states: value zero (+/-).

**Averaging:**

- Whenever two strong nodes meet, they average values

**Conquer:**

- **Strong** nodes sway **weak** nodes towards their decision.



# AVC in Action

Initially:  $+m$  or  $-m$ , **odd** integers

**Strong** states: non-zero absolute value.

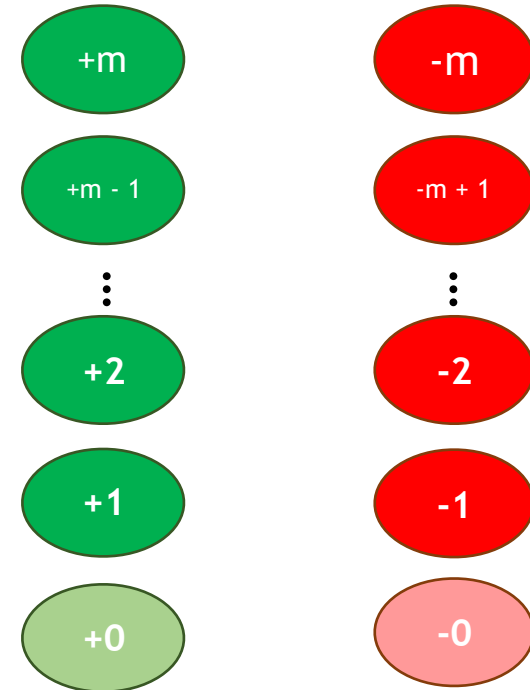
**Weak** states: value zero (+/-).

**Averaging:**

- Whenever two strong nodes meet, they average values

**Conquer:**

- **Strong** nodes sway **weak** nodes towards their decision.



# AVC in Action

Initially:  $+m$  or  $-m$ , **odd** integers

**Strong** states: non-zero absolute value.

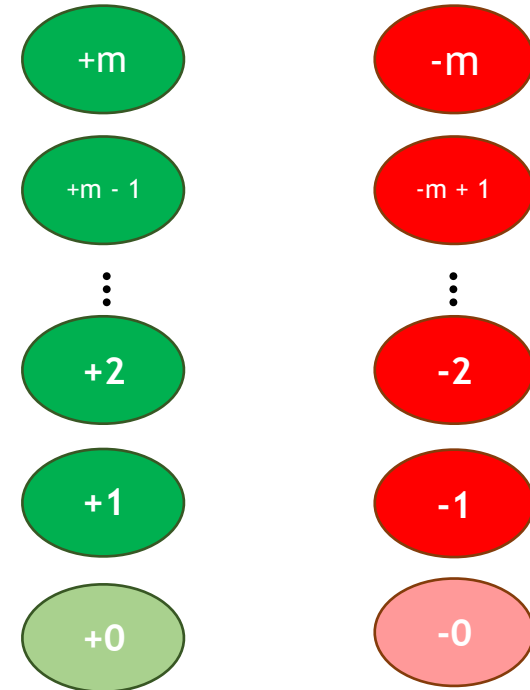
**Weak** states: value zero (+/-).

**Averaging:**

- Whenever two strong nodes meet, they average values

**Conquer:**

- **Strong** nodes sway **weak** nodes towards their decision.



**Note:** For  $m = 1$ , we obtain a variant of 4EM.

# AVC in Action

Initially:  $+m$  or  $-m$ , odd integers

Strong states: non-zero absolute value.

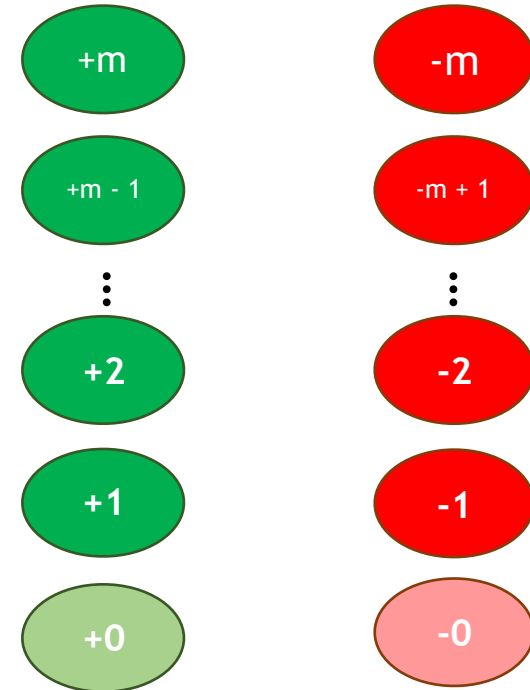
Weak states: value zero (+/-).

Averaging:

- Whenever two strong nodes meet, they average values

Conquer:

- Strong nodes sway weak nodes towards their decision.



**Note:** For  $m = 1$ , we obtain a variant of 4EM.

**Disclaimer:** original protocol is more complicated for technical reasons

Summing up

# Summing up

Theorem 1 [AGV15]: Given fixed  $m < n$ , AVC solves majority exactly in expected parallel time  $O(\log n / (m \epsilon) + \log n \log m)$ , using  $s = O(m + \log n \log m)$  total states.



# Summing up

Theorem 1 [AGV15]: Given fixed  $m < n$ , AVC solves majority exactly in expected parallel time  $O(\log n / (m \epsilon) + \log n \log m)$ , using  $s = O(m + \log n \log m)$  total states.

- In short:
  - If  $m \approx 1 / \epsilon$ , then running time is always **poly-logarithmic**
  - If  $\epsilon = 1 / n$ , then  $m$  needs to be **linear** in  $n$
  - $10^{23}$  molecules  $\rightarrow O(10^{23})$  states?!
  
- $10^{23}$  molecules  $\rightarrow O(2^{32})$  states )
- The idea: **quantize** integer states to powers of two

# Summing up

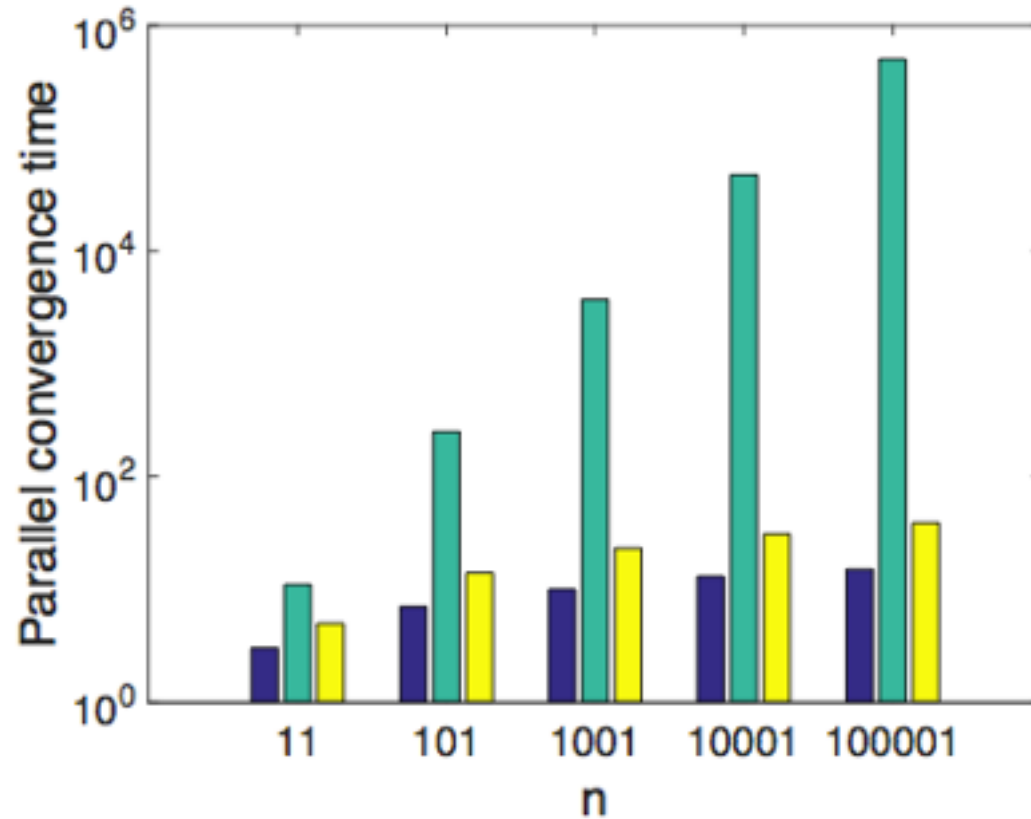
Theorem 1 [AGV15]: Given fixed  $m < n$ , AVC solves majority exactly in expected parallel time  $O(\log n / (m \epsilon) + \log n \log m)$ , using  $s = O(m + \log n \log m)$  total states.

- In short:
  - If  $m \approx 1 / \epsilon$ , then running time is always **poly-logarithmic**
  - If  $\epsilon = 1 / n$ , then  $m$  needs to be **linear** in  $n$
  - $10^{23}$  molecules  $\rightarrow O(10^{23})$  states?!

Theorem 2 [AAEGR16]: **logAVC** solves majority exactly in expected parallel time  $O(\log^3 n)$ , using  $s = O(\log^2 n)$  total states.

- $10^{23}$  molecules  $\rightarrow O(2^{32})$  states )
- The idea: **quantize** integer states to powers of two

# Is AVC any good?



Results are for  $\epsilon = O(1 / n)$

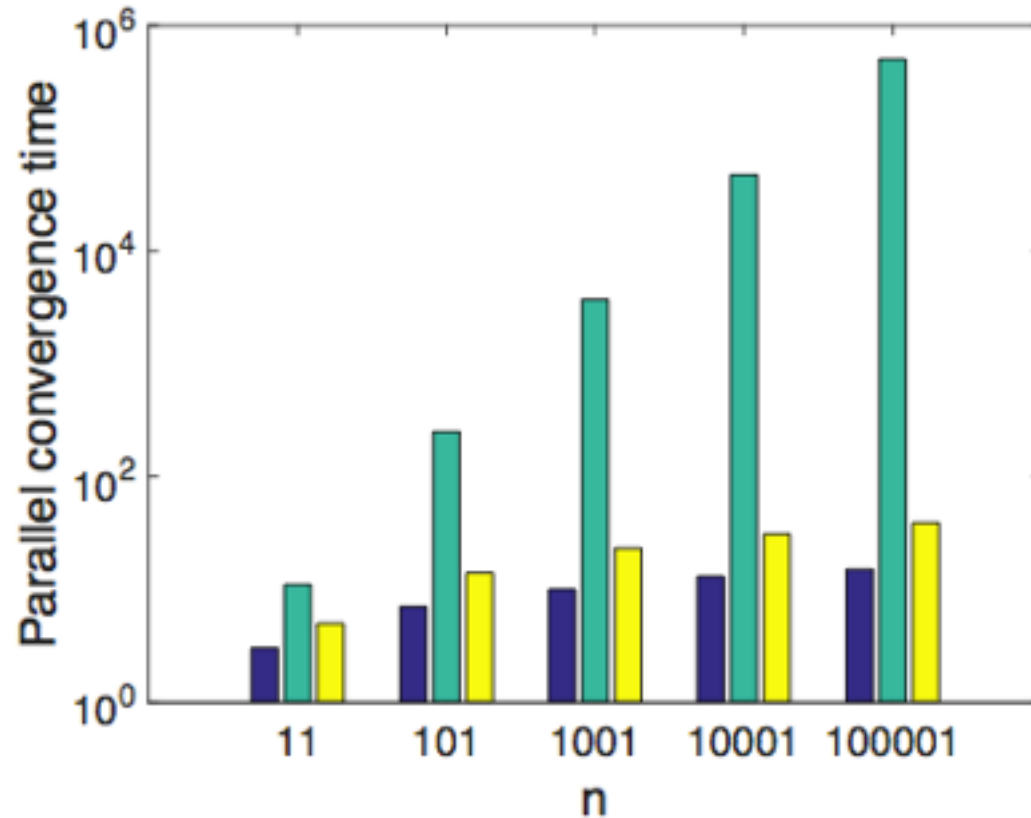
Legend:

Blue = 3AM

Green = 4EM

Yellow = AVC / logAVC

# Is AVC any good?



Results are for  $\epsilon = O(1 / n)$

Legend:

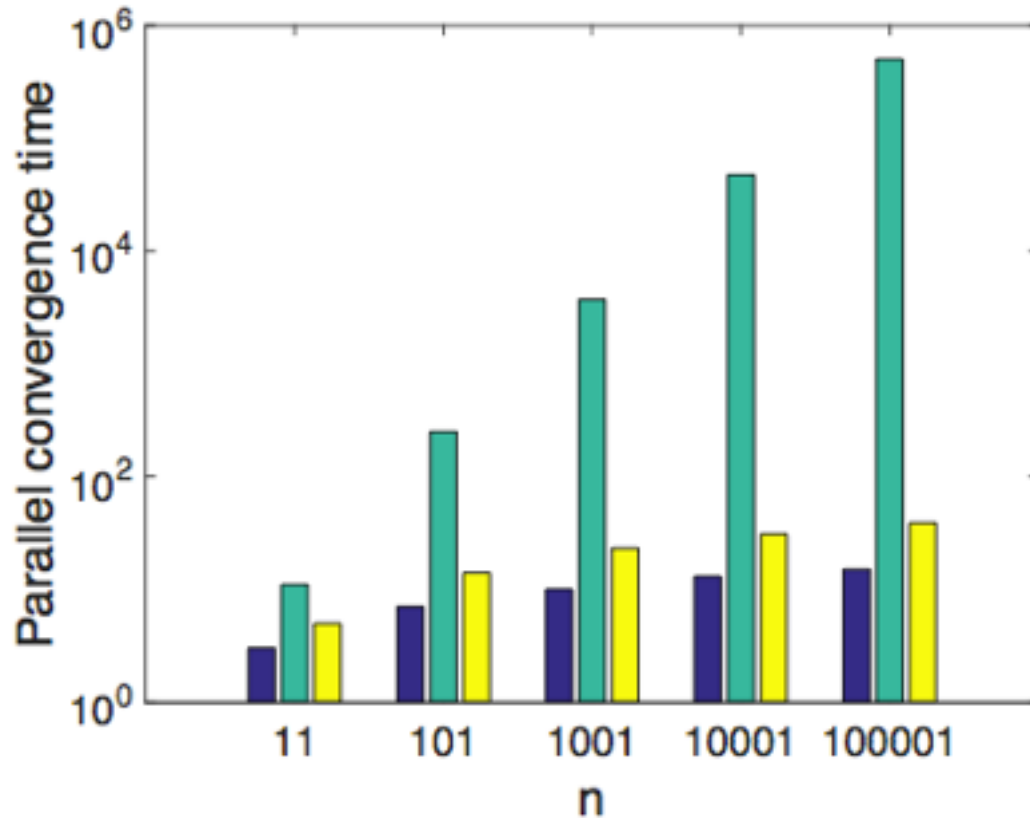
Blue = 3AM

Green = 4EM

Yellow = AVC / logAVC

Is AVC implementable?

# Is AVC any good?



Results are for  $\epsilon = O(1 / n)$

Legend:

Blue = 3AM

Green = 4EM

Yellow = AVC / logAVC

Is AVC implementable?

Challenging: currently, small constant number of states implementable.

# Time-Space Trade-Offs

# Time-Space Trade-Offs

Theorem A: Any protocol using  $s < \frac{1}{2} \log \log n$  states per node and solving majority with **discrepancy**  $\epsilon$  must have expected stabilization time  $> n / (2^s + \epsilon n)^2$ .

# Time-Space Trade-Offs

Theorem A: Any protocol using  $s < \frac{1}{2} \log \log n$  states per node and solving majority with **discrepancy**  $\epsilon$  must have expected stabilization time  $> n / (2^s + \epsilon n)^2$ .

- In particular:
  - If  $s = \text{constant}$  and  $\epsilon n = \text{constant}$ , then stabilization time **linear in  $n$**
  - If  $s = O(\log \log n)$  and  $\epsilon n = \text{constant}$ , then stabilization time  $> n / \text{polylog } n$



# Time-Space Trade-Offs

Theorem A: Any protocol using  $s < \frac{1}{2} \log \log n$  states per node and solving majority with **discrepancy**  $\epsilon$  must have expected stabilization time  $> n / (2^s + \epsilon n)^2$ .

- In particular:
  - If  $s = \text{constant}$  and  $\epsilon n = \text{constant}$ , then stabilization time **linear in  $n$**
  - If  $s = O(\log \log n)$  and  $\epsilon n = \text{constant}$ , then stabilization time  $> n / \text{polylog } n$

**Complex molecules** are needed for **deterministic computation**.

# Discussion

# Discussion

**Molecular computation is fertile ground  
for algorithmic research.**

# Discussion

Molecular computation is fertile ground for algorithmic research.

There are **inherent space-time trade-offs** when designing **deterministic** population protocols.

# Discussion

Molecular computation is fertile ground for algorithmic research.

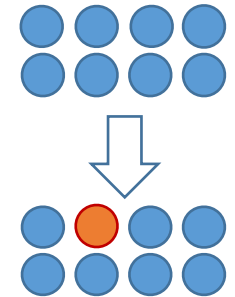
There are **inherent space-time trade-offs** when designing **deterministic** population protocols.

## Open Challenges:

- Tighter trade-off bounds
- Other problems: **plurality**, **approximate counting**
- **Modeling faulty interactions** (leaks)
- **Large-scale simulation** of molecular algorithms

# Leader Election

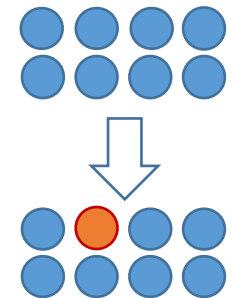
- **Input:** All nodes start in the same initial state
- **Output:**



Algorithm	Number of States	Convergence Time
Trivial Leader Election	2	$\Omega(n^2)$
Leader-Minion [AG, ICALP 2015]	$O(\log^3 n)$	$O(\log^3 n)$
Lottery Leader Election [AAEGR16]	$O(\log^2 n)$	$O(\log^{5.3} n \log \log n)$

# Leader Election

- **Input:** All nodes start in the same initial state
- **Output:**
  - Exactly one node is in a “leader” state, remains leader forever



Algorithm	Number of States	Convergence Time
Trivial Leader Election	2	$\Omega(n^2)$
Leader-Minion [AG, ICALP 2015]	$O(\log^3 n)$	$O(\log^3 n)$
Lottery Leader Election [AAEGR16]	$O(\log^2 n)$	$O(\log^{5.3} n \log \log n)$

# The Impossibility Result

Theorem A: Any protocol using  $< \frac{1}{2} \log \log n$  states per node and electing  $L$  leaders will have expected stabilization time  $> n / (C \text{polylog } n L^2)$ .



# The Impossibility Result

Theorem A: Any protocol using  $< \frac{1}{2} \log \log n$  states per node and electing  $L$  leaders will have expected stabilization time  $> n / (C \text{polylog } n L^2)$ .

- Example:
  - $O(\log \log n)$  states / node, one leader
  - Stabilization time  $> n / \text{polylog } n$  (quasi-linear)
  - Generalizes a recent result by Doty and Soloveichik [DISC15] to super-constant states



# Bonus: A Cute Algorithm

- The goal: approximate  $n$
- The state:
  - A flip bit  $F$ , initially 0
  - A counter “variable”  $C$ , initially 0
- The algorithm:
  - Stage 1: do four interactions, updating  $F = 1 - F'$
  - Stage 2: increment counter  $C$  until you first see  $F' = 1$
  - Stage 3: exchange  $C$  with interaction partner, setting  $C = \max(C, C')$
- The guarantee:
  - The convergence value is  $(1 - \text{eps}) \log n < C < (1 + \text{eps}) \log n$ , with high probability