

Polynomial Anonymous Dynamic Distributed Computing without a Unique Leader

Dariusz R. Kowalski
U. Liverpool (UK)

Miguel A. Mosteiro
Pace Univ. (USA)

ICALP 2019

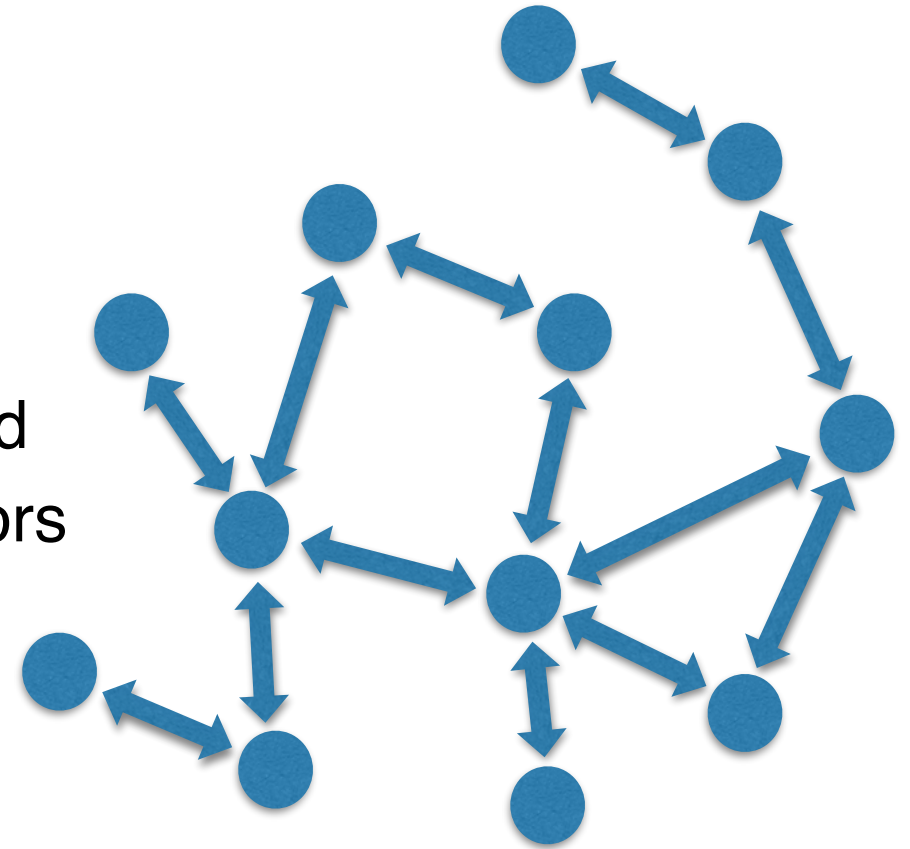
Anonymous Dynamic Networks

- **Fixed set of n nodes**
 - No identifiers or labels
- **Synchronous communication** : At each round
 - a node broadcasts a message to its neighbors
 - receives the messages of its neighbors
 - executes some local computation
- **1-interval connectivity [1]**
 - communication links may change from round to round, but
 - at each round the network is connected



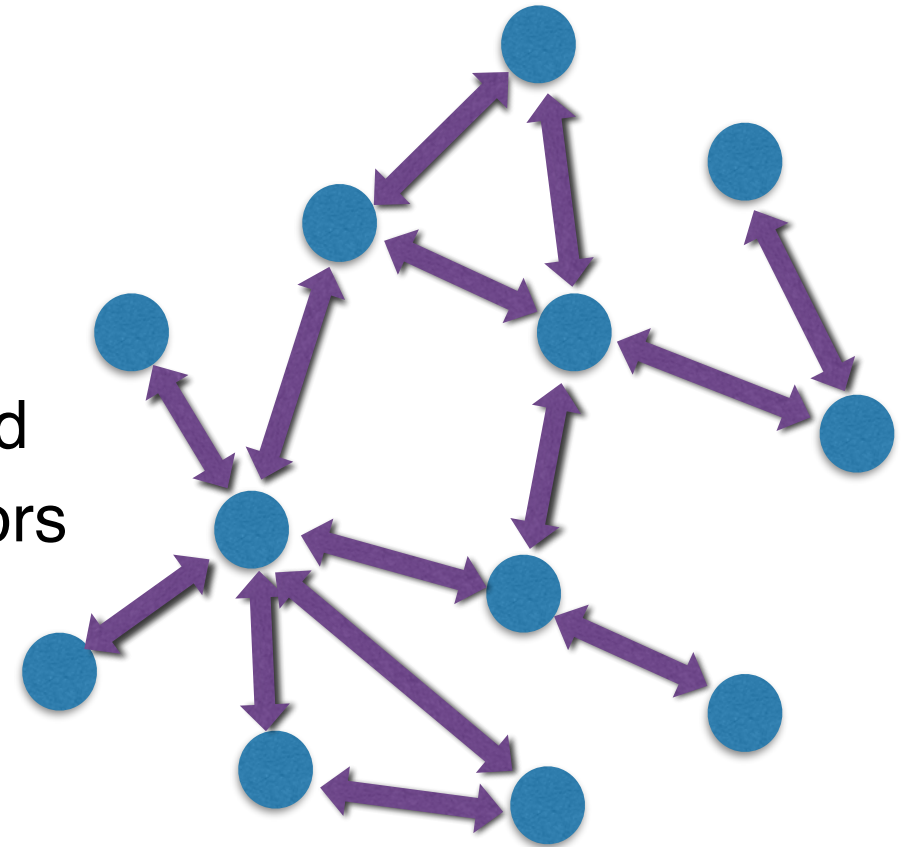
Anonymous Dynamic Networks

- **Fixed set of n nodes**
 - No identifiers or labels
- **Synchronous communication** : At each round
 - a node broadcasts a message to its neighbors
 - receives the messages of its neighbors
 - executes some local computation
- **1-interval connectivity [1]**
 - communication links may change from round to round, but
 - at each round the network is connected



Anonymous Dynamic Networks

- **Fixed set of n nodes**
 - No identifiers or labels
- **Synchronous communication** : At each round
 - a node broadcasts a message to its neighbors
 - receives the messages of its neighbors
 - executes some local computation
- **1-interval connectivity [1]**
 - communication links may change from round to round, but
 - at each round the network is connected



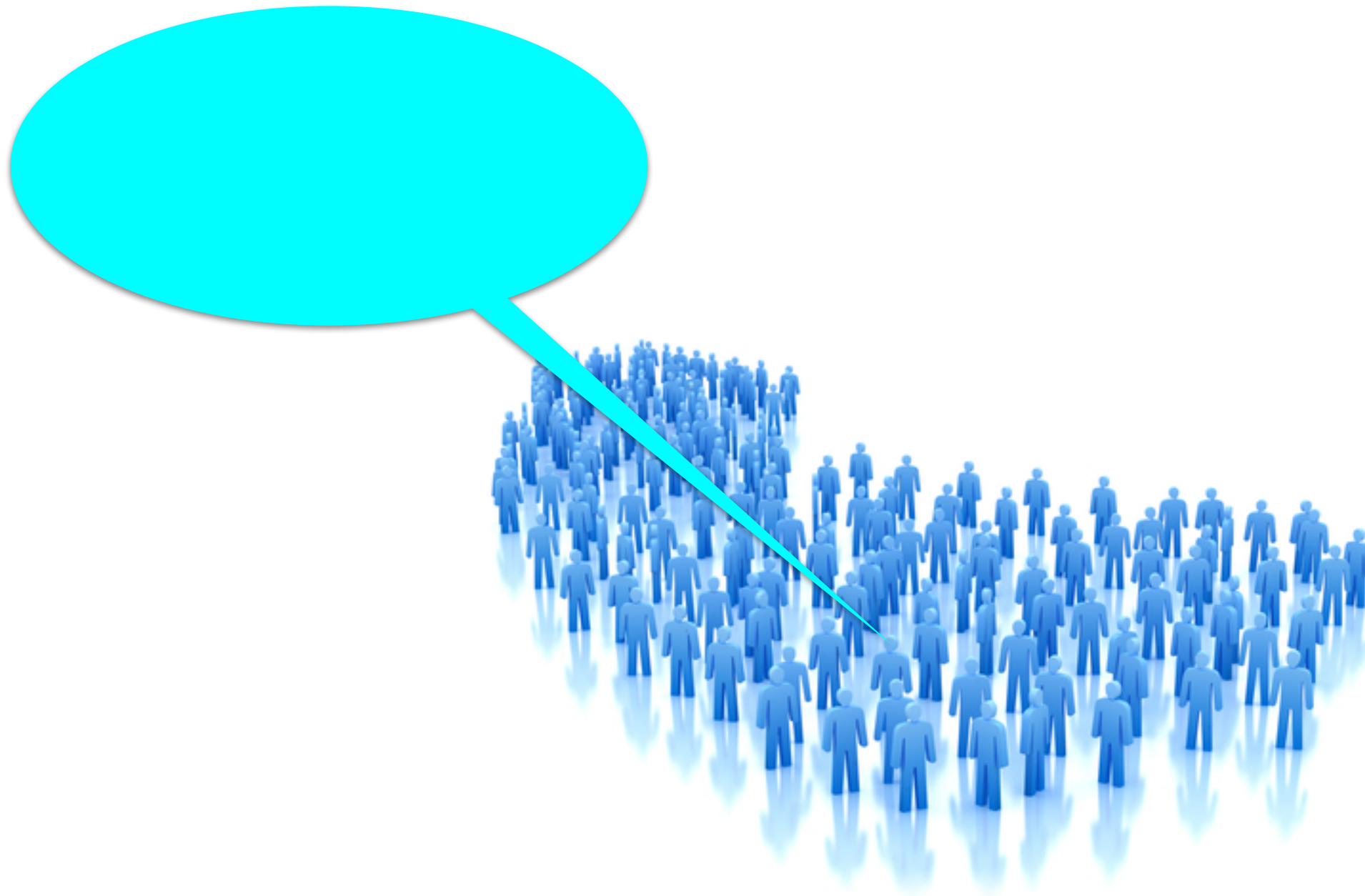
The Counting Problem

How do you count the size of your group,
if the members are all identical and move?



The Counting Problem

How do you count the size of your group,
if the members are all identical and move?



The Counting Problem

How do you count the size of your group,
if the members are all identical and move?

You all look the same,
did I already count you?



The Counting Problem

How do you count the size of your group,
if the members are all identical and move?

You all look the same,
did I already count you?



The Counting Problem

How do you count the size of your group,
if the members are all identical and move?

You all look the same,
did I already count you?



The Counting Problem

How do you count the size of your group,
if the members are all identical and move?

You all look the same,
did I already count you?

I don't know!
You also look the same as
everyone else!!



Why do we care?

The problem is clean, but why do we care?



Why do we care?

The problem is clean, but why do we care?

Distributed algorithms

need the number of processors to decide **termination**.

We need a **protocol**: « Given a system of **n** nodes,
all nodes eventually terminate knowing **n** ».



Why do we care?



Vehicle, asset, person & pet
monitoring & controlling



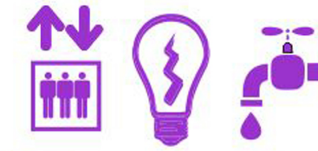
Agriculture automation



Energy consumption



Security &
surveillance



Building management



Embedded
Mobile

Internet of things

Everyday things
get connected for smarter
tomorrow



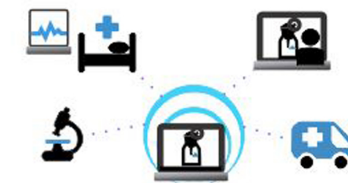
M2M & wireless
sensor network



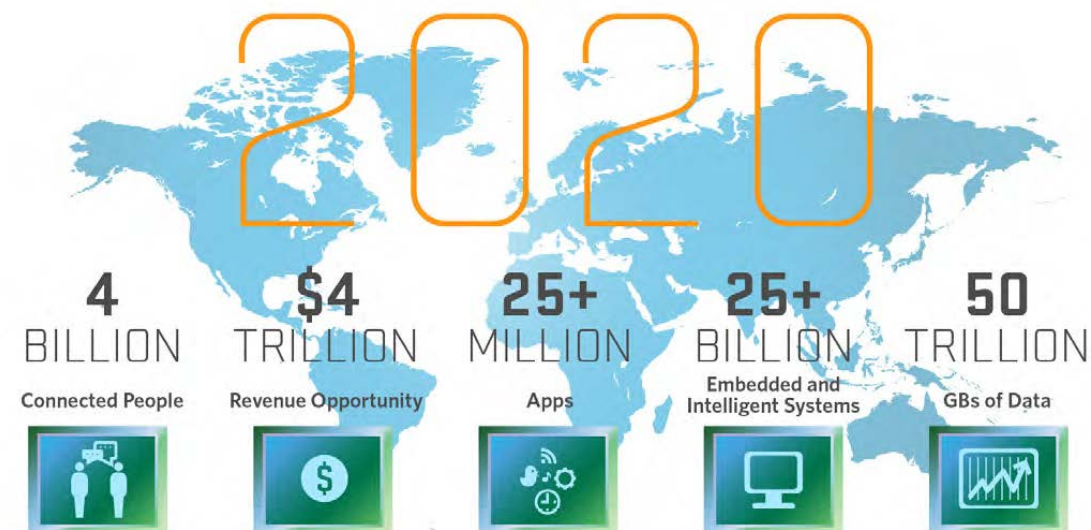
Everyday things



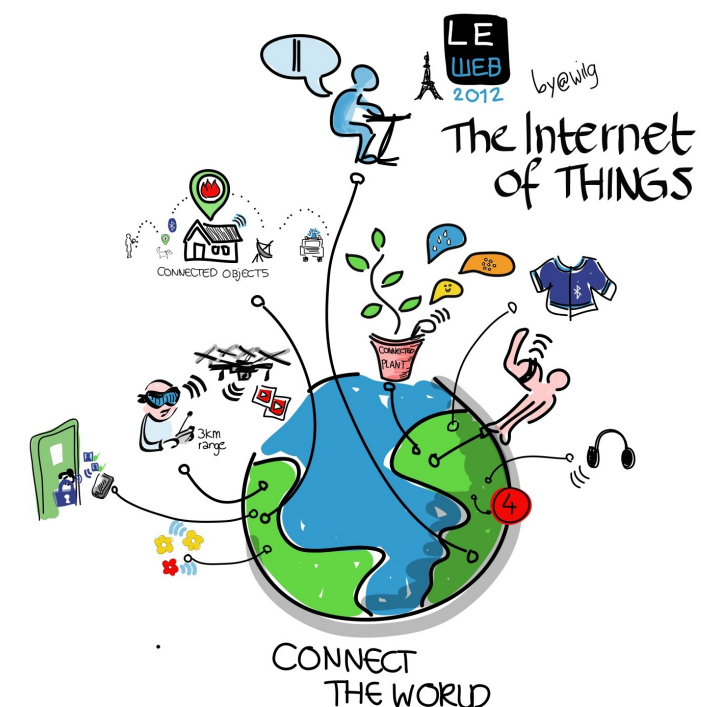
Smart homes & cities



Telemedicine & healthcare



Source: Mario Morales, IDC



Why do we care?



Vehicle, asset, person & pet
monitoring & controlling



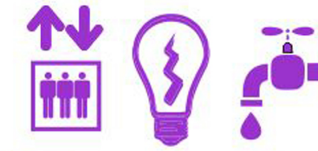
Agriculture automation



Energy consumption



Security &
surveillance



Building management



Embedded
Mobile

Internet of things

Everyday things
get connected for smarter
tomorrow



M2M & wireless
sensor network



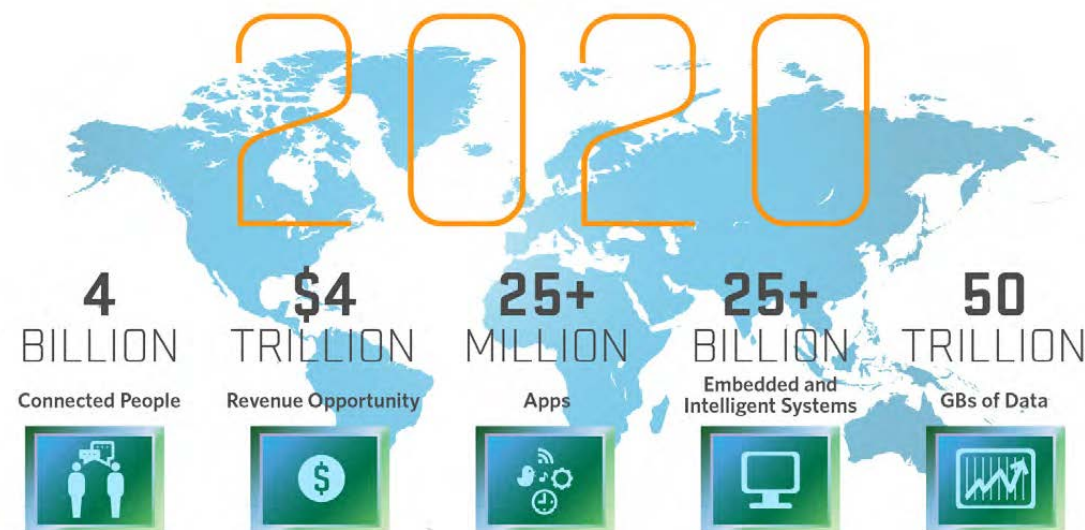
Everyday things



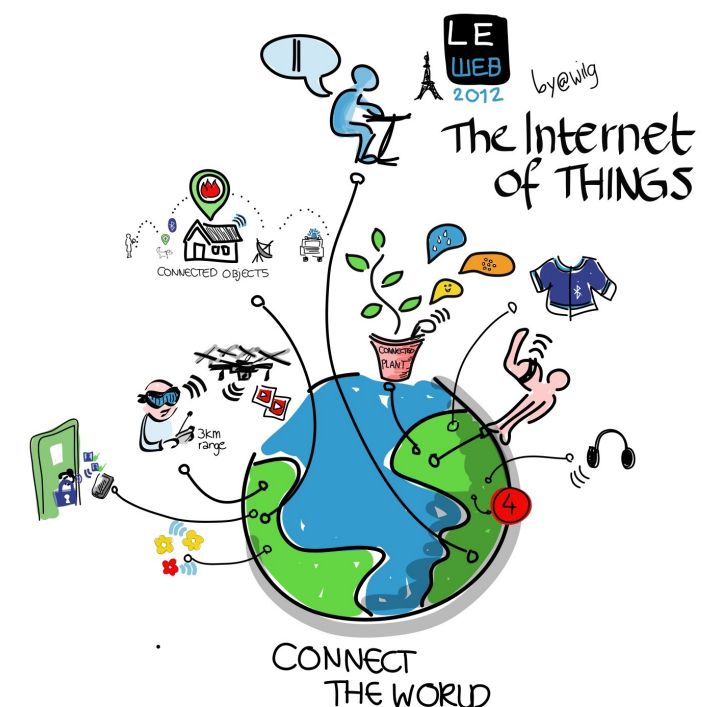
Smart homes & cities



Telemedicine & healthcare



Source: Mario Morales, IDC



Previous work

algorithm	needs			computes	stops?	complexity	
	distinguished nodes	size upper bound N	dynamic maximum degree u.b. d_{\max}			time	space
<i>Degree Counting</i> [5]	1		✓	$O(d_{\max}^n)$	✓	$O(n)$	
<i>Conscious</i> [2]	1	✓	✓	n	✓	$O(e^{N^2} N^3) \Rightarrow O(e^{d_{\max}^{2n}} d_{\max}^{3n})$ using [5]	
<i>Unconscious</i> [2]	1			n	No	No theoretical bounds	
\mathcal{A}_{OP} [3]	1		Oracle for each node	n	Eventually	Unknown	
EXT [1]	1			n	✓	$O(n^{n+4})$	EXPSPACE
<i>Incremental Counting</i> [6]	1		✓	n	✓	$O\left(n (2d_{\max})^{n+1} \frac{\ln n}{\ln d_{\max}}\right)$	
<i>Methodical Counting</i> [4]	1			n	✓	$O(n^5 \ln^2 n)$	PSPACE

restrictions/
shortcomings

[5] O. Michail, I. Chatzigiannakis, and P. G. Spirakis. Naming and counting in anonymous unknown dynamic networks. SSS 2013.

[2] G. A. Di Luna, R. Baldoni, S. Bonomi, and I. Chatzigiannakis. Conscious and unconscious counting on anonymous dynamic networks. ICDCN 2014.

[3] G. A. Di Luna, R. Baldoni, S. Bonomi, and I. Chatzigiannakis. Counting in anonymous dynamic networks under worst-case adversary. ICDCS 2014.

[6] A. Milani and M. A. Mosteiro. A faster counting protocol for anonymous dynamic networks. OPODIS 2015.

[1] R. Baldoni and G. A. Di Luna. Non trivial computations in anonymous dynamic networks. OPODIS 2015.

[4] D. Kowalski and M. A. Mosteiro. Polynomial counting in anonymous dynamic networks with applications to anonymous dynamic algebraic computations. ICALP 2018.

Previous work

algorithm	needs			computes	stops?	complexity	
	distinguished nodes	size upper bound N	dynamic maximum degree u.b. d_{\max}			time	space
<i>Degree Counting</i> [5]	1		✓	$O(d_{\max}^n)$	✓	$O(n)$	
<i>Conscious</i> [2]	1	✓	✓	n	✓	$O(e^{N^2} N^3) \Rightarrow O(e^{d_{\max}^{2n}} d_{\max}^{3n})$ using [5]	
<i>Unconscious</i> [2]	1			n	No	No theoretical bounds	
\mathcal{A}_{OP} [3]	1		Oracle for each node	n	Eventually	Unknown	
EXT [1]	1			n	✓	$O(n^{n+4})$	EXPSPACE
<i>Incremental Counting</i> [6]	1		✓	n	✓	$O\left(n (2d_{\max})^{n+1} \frac{\ln n}{\ln d_{\max}}\right)$	
<i>Methodical Counting</i> [4]	1			n	✓	$O(n^5 \ln^2 n)$	PSPACE

restrictions/
shortcomings

first
polynomial

[5] O. Michail, I. Chatzigiannakis, and P. G. Spirakis. Naming and counting in anonymous unknown dynamic networks. SSS 2013.

[2] G. A. Di Luna, R. Baldoni, S. Bonomi, and I. Chatzigiannakis. Conscious and unconscious counting on anonymous dynamic networks. ICDCN 2014.

[3] G. A. Di Luna, R. Baldoni, S. Bonomi, and I. Chatzigiannakis. Counting in anonymous dynamic networks under worst-case adversary. ICDCS 2014.

[6] A. Milani and M. A. Mosteiro. A faster counting protocol for anonymous dynamic networks. OPODIS 2015.

[1] R. Baldoni and G. A. Di Luna. Non trivial computations in anonymous dynamic networks. OPODIS 2015.

[4] D. Kowalski and M. A. Mosteiro. Polynomial counting in anonymous dynamic networks with applications to anonymous dynamic algebraic computations. ICALP 2018.

Previous work

algorithm	needs			computes	stops?	complexity	
	distinguished nodes	size upper bound N	dynamic maximum degree u.b. d_{\max}			time	space
<i>Degree Counting</i> [5]	1		✓	$O(d_{\max}^n)$	✓	$O(n)$	
<i>Conscious</i> [2]	1	✓	✓	n	✓	$O(e^{N^2} N^3) \Rightarrow O(e^{d_{\max}^{2n}} d_{\max}^{3n})$ using [5]	
<i>Unconscious</i> [2]	1			n	No	No theoretical bounds	
\mathcal{A}_{OP} [3]	1		Oracle for each node	n	Eventually	Unknown	
EXT [1]	1			n	✓	$O(n^{n+4})$	EXPSPACE
<i>Incremental Counting</i> [6]	1		✓	n	✓	$O\left(n (2d_{\max})^{n+1} \frac{\ln n}{\ln d_{\max}}\right)$	
<i>Methodical Counting</i> [4]	1			n	✓	$O(n^5 \ln^2 n)$	PSPACE

restrictions/
shortcomings

needs at
least one [5]

first
polynomial

[5] O. Michail, I. Chatzigiannakis, and P. G. Spirakis. Naming and counting in anonymous unknown dynamic networks. SSS 2013.

[2] G. A. Di Luna, R. Baldoni, S. Bonomi, and I. Chatzigiannakis. Conscious and unconscious counting on anonymous dynamic networks. ICDCN 2014.

[3] G. A. Di Luna, R. Baldoni, S. Bonomi, and I. Chatzigiannakis. Counting in anonymous dynamic networks under worst-case adversary. ICDCS 2014.

[6] A. Milani and M. A. Mosteiro. A faster counting protocol for anonymous dynamic networks. OPODIS 2015.

[1] R. Baldoni and G. A. Di Luna. Non trivial computations in anonymous dynamic networks. OPODIS 2015.

[4] D. Kowalski and M. A. Mosteiro. Polynomial counting in anonymous dynamic networks with applications to anonymous dynamic algebraic computations. ICALP 2018.

Questions

- Can we count deterministically with **more than one special node**?
- **What information** about the special nodes is needed?
- **How “special”?** Can the nodes be identical and just have two different programs?
- Can we let the nodes choose program at *random* and **make them all identical**?
- Can we **tighten previous bounds**?

Questions

- Can we count deterministically with **more than one special node**? **Yes!**
- **What information** about the special nodes is needed? **Count**
- **How “special”?** Can the nodes be identical and just have two different programs? **Yes!**
- Can we let the nodes choose program at *random* and **make them all identical**? **Yes!**
- Can we **tighten previous bounds**? **Yes!**

Results

ℓ black nodes and $n-\ell$ white nodes:

- Impossibility:
 - Deterministic counting: not possible if ℓ is unknown
 - Randomized counting: if ℓ is unknown or zero, exist executions that do not stop

Results

ℓ black nodes and $n-\ell$ white nodes:

- Impossibility:
 - Deterministic counting: not possible if ℓ is unknown
 - Randomized counting: if ℓ is unknown or zero, exist executions that do not stop
- Methodical Multi-Counting (MMC) algorithm:
 - *all* nodes obtain n and terminate
 - no network info needed except ℓ
- Leader-less Methodical Counting (LLMC) algorithm:
 - first algorithm applicable to ADNs with all identical nodes

Results

algorithm	needs			computes	stops?	complexity	
	distinguished nodes	size upper bound N	dynamic maximum degree u.b. d_{\max}			time	space
<i>Degree Counting</i> [5]	1		✓	$O(d_{\max}^n)$	✓	$O(n)$	
<i>Conscious</i> [2]	1	✓	✓	n	✓	$O(e^{N^2} N^3) \Rightarrow O(e^{d_{\max}^{2n}} d_{\max}^{3n})$ using [5]	
<i>Unconscious</i> [2]	1			n	No	No theoretical bounds	
\mathcal{A}_{OP} [3]	1		Oracle for each node	n	Eventually	Unknown	
EXT [1]	1			n	✓	$O(n^{n+4})$	EXPSPACE
<i>Incremental Counting</i> [6]	1		✓	n	✓	$O\left(n (2d_{\max})^{n+1} \frac{\ln n}{\ln d_{\max}}\right)$	
<i>Methodical Counting</i> [4]	1			n	✓	$O(n^5 \ln^2 n)$	PSPACE
METHODICAL MULTI-COUNTING [This work]	$\ell \geq 1$			n	✓	$O((n^{4+\epsilon}/\ell) \log^3 n)$ for any $\epsilon > 0$	PSPACE
LEADER-LESS METHODICAL-COUNTING [This work]	0			n prob. $\geq 1 - \zeta$ for any $\zeta > 0$	✓	$O(\eta^{4+\epsilon} \log^3 \eta)$ for any $\epsilon > 0$ and $\eta = \max\{n, \lceil 12/\zeta \rceil\}$	PSPACE

Results

algorithm	needs			computes	stops?	complexity	
	distinguished nodes	size upper bound N	dynamic maximum degree u.b. d_{\max}			time	space
<i>Degree Counting</i> [5]	1		✓	$O(d_{\max}^n)$	✓	$O(n)$	
<i>Conscious</i> [2]	1	✓	✓	n	✓	$O(e^{N^2} N^3) \Rightarrow O(e^{d_{\max}^{2n}} d_{\max}^{3n})$ using [5]	
<i>Unconscious</i> [2]	1			n	No	No theoretical bounds	
\mathcal{A}_{OP} [3]	1		Oracle for each node	n	Eventually	Unknown	
EXT [1]	1			n	✓	$O(n^{n+4})$	EXPSPACE
<i>Incremental Counting</i> [6]	1		✓	n	✓	$O\left(n (2d_{\max})^{n+1} \frac{\ln n}{\ln d_{\max}}\right)$	
<i>Methodical Counting</i> [4]	1			n	✓	$O(n^5 \ln^2 n)$	PSPACE
METHODICAL MULTI-COUNTING [This work]	$\ell \geq 1$			n	✓	$O((n^{4+\epsilon}/\ell) \log^3 n)$ for any $\epsilon > 0$	PSPACE
LEADER-LESS METHODICAL-COUNTING [This work]	0			n prob. $\geq 1 - \zeta$ for any $\zeta > 0$	✓	$O(\eta^{4+\epsilon} \log^3 \eta)$ for any $\epsilon > 0$ and $\eta = \max\{n, \lceil 12/\zeta \rceil\}$	PSPACE

first with $\ell \neq 1$

$\sim n\ell / \log n$ speedup (for $\zeta \in \Omega(1/n)$),
faster than MC even for $\ell = 1$

Deterministic Counting

Even knowing ℓ , trivial application of ℓ instances of MC not clear:

- How the black nodes communicate?
- How do they compare/combine final results?
- Black nodes are all identical!

Deterministic Counting

Even knowing ℓ , trivial application of ℓ instances of MC not clear:

- How the black nodes communicate?
- How do they compare/combine final results?
- Black nodes are all identical!

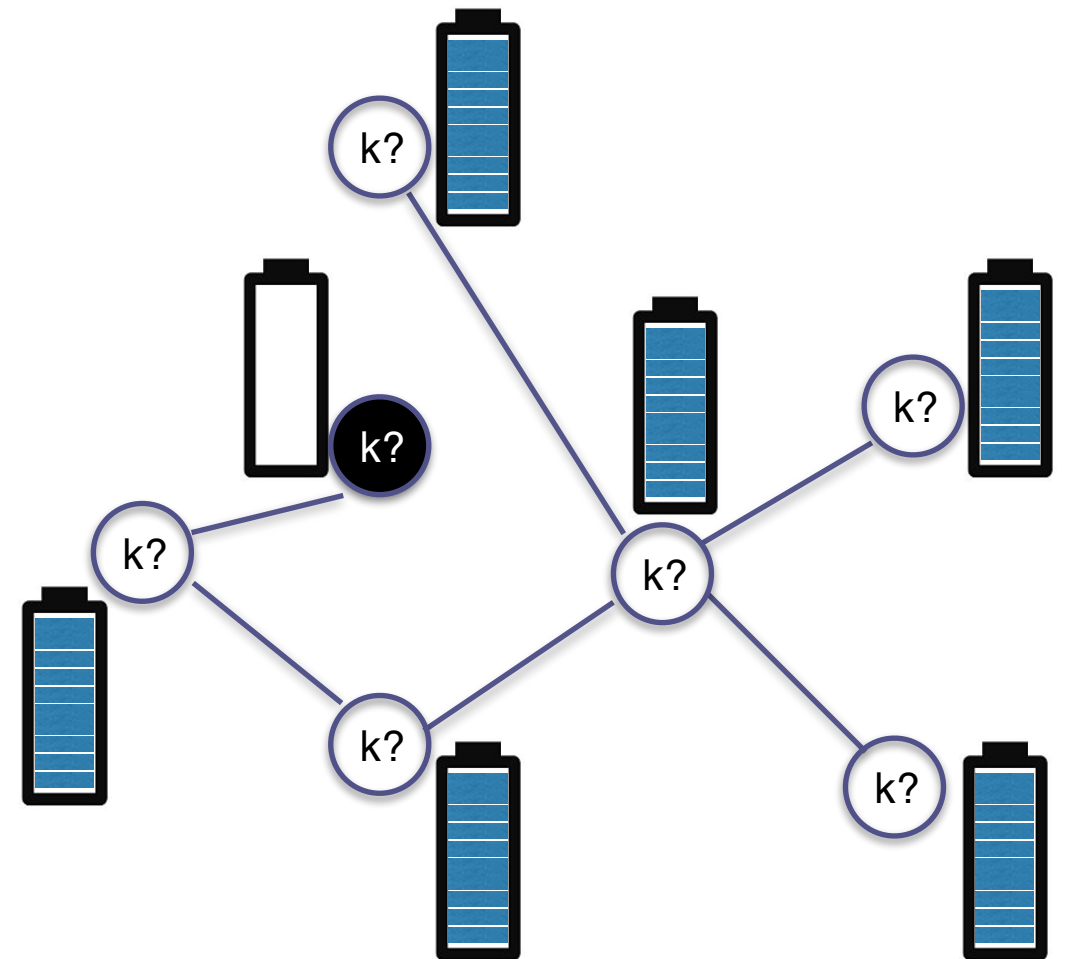
MMC Key Ingredients:

- try network size estimates $k = \ell + 1, 2(\ell + 1), 4(\ell + 1), \dots$ binary search after estimate $k > n$
- share some potential values iteratively
- all nodes (black and white) share potential
- black nodes remove potential from the system every now and then
- carefully designed alarms allow to detect correct or wrong estimate

MMC Structure

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$



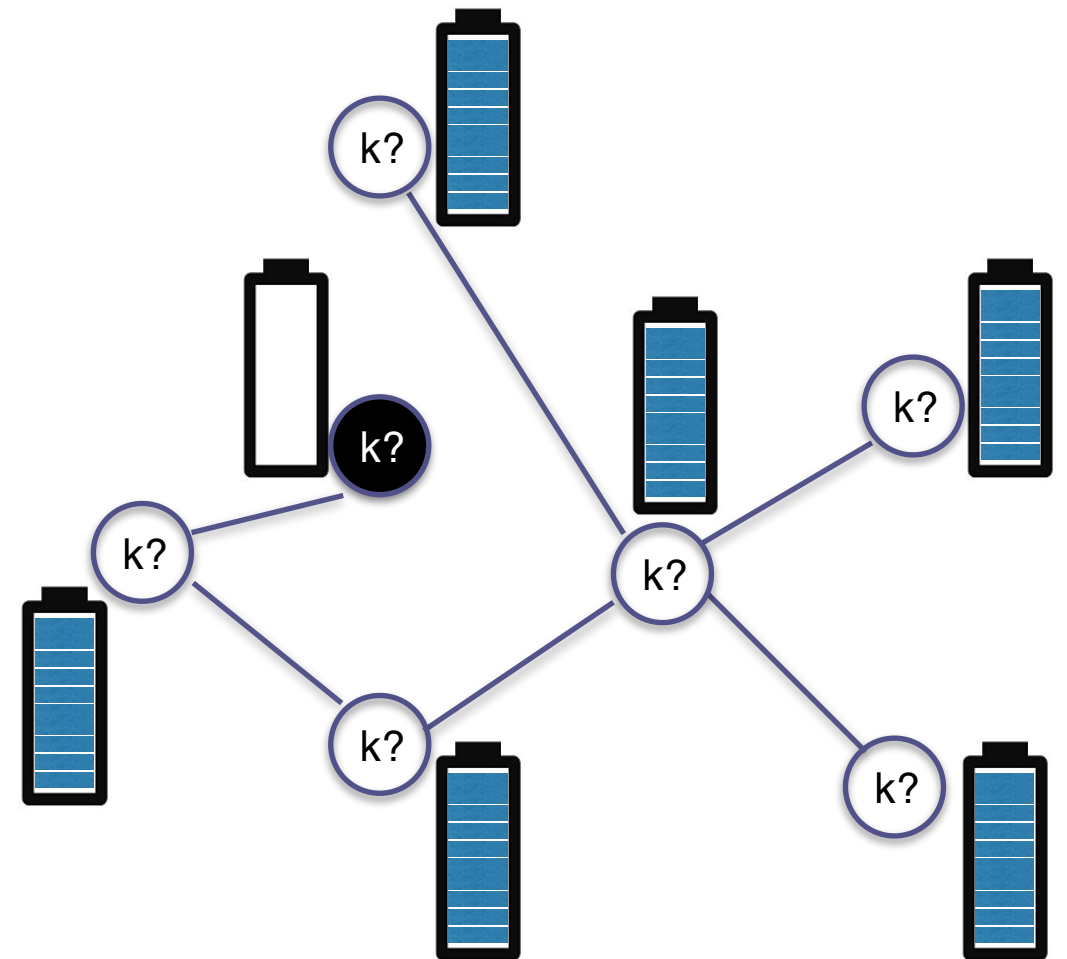
MMC Structure

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

(to let blacks remove “enough” potential ρ)



MMC Structure

epochs:

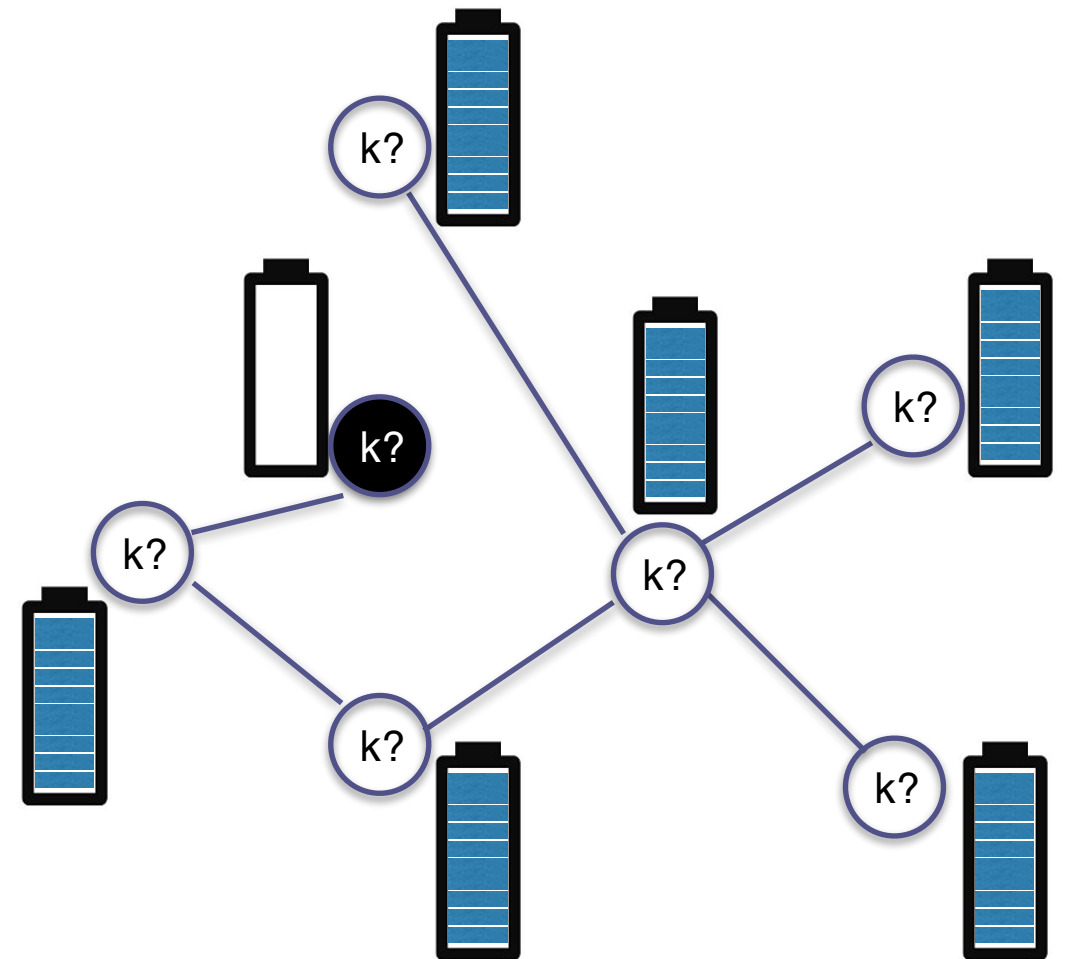
- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)



epochs:

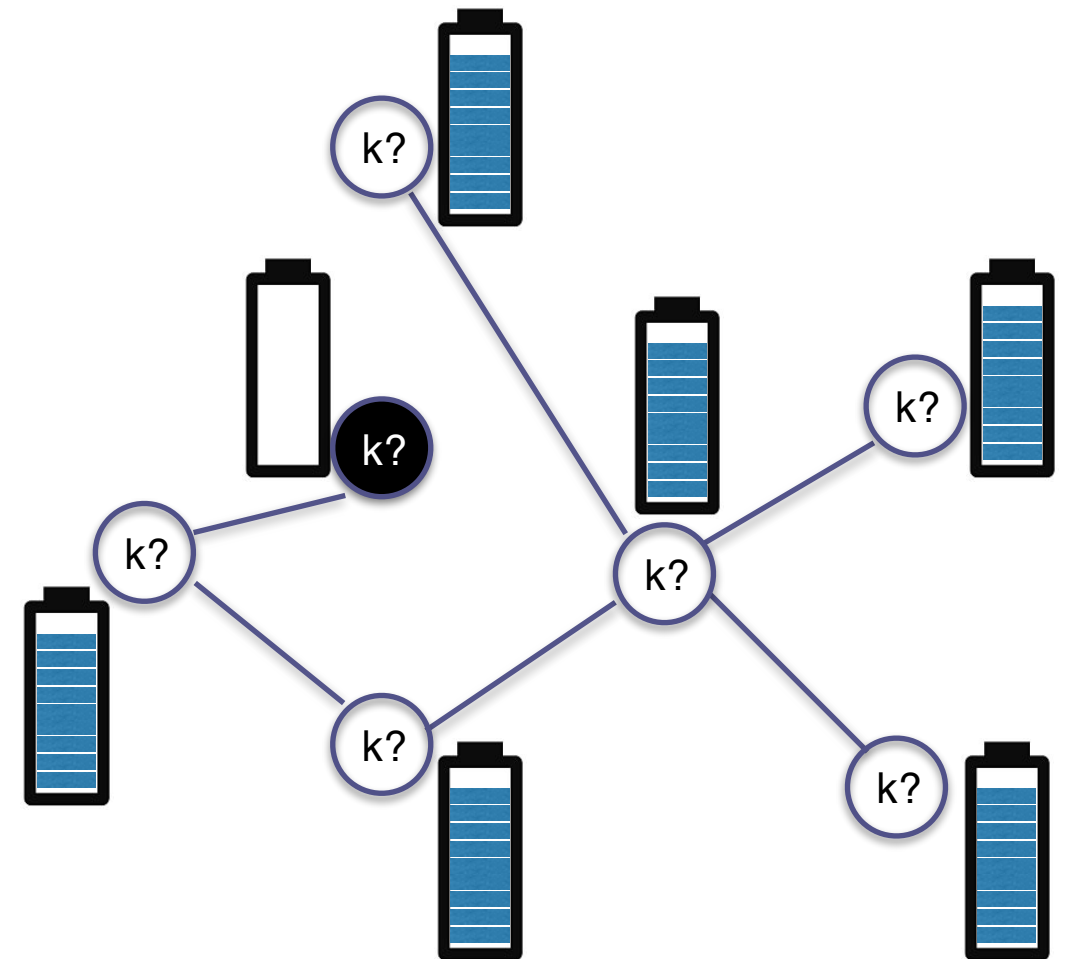
- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)



$\rho =$



MMC Epoch Example

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

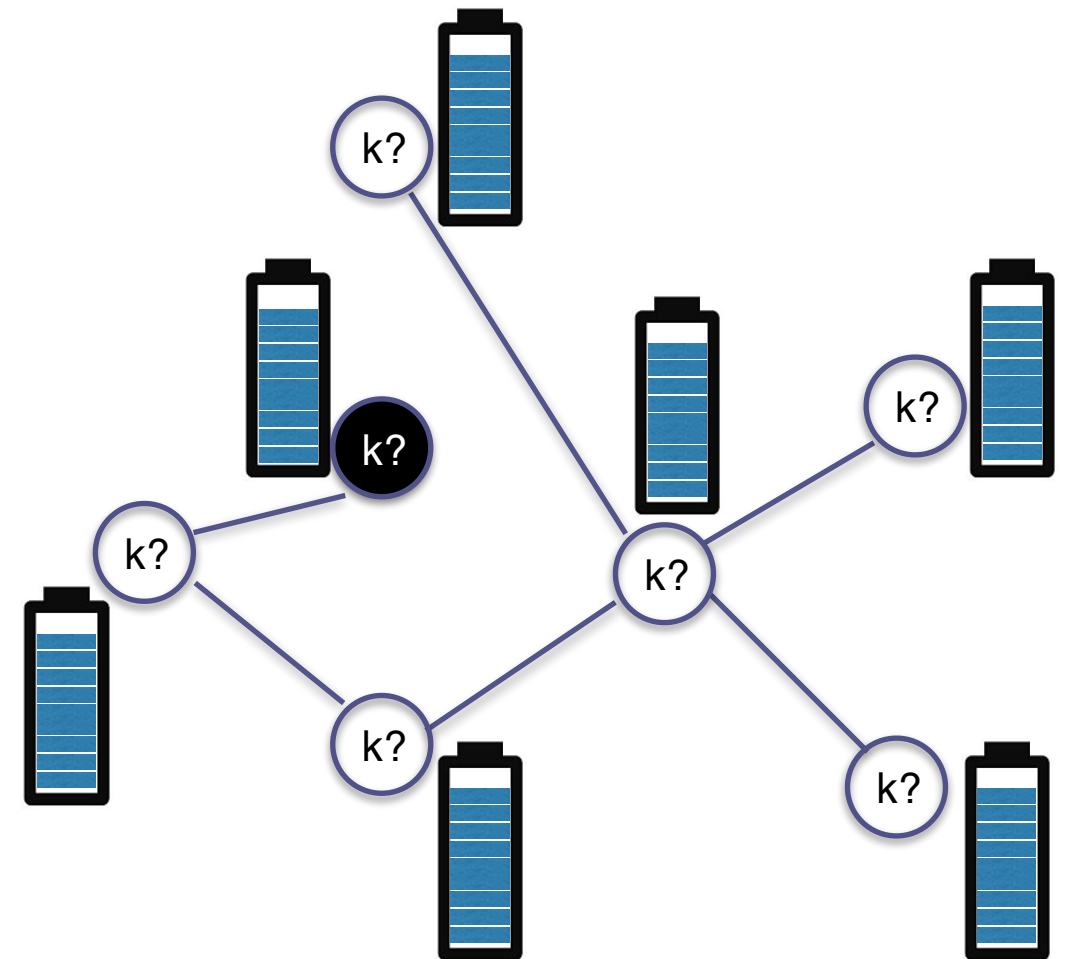
(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)

mass distribution:

- broadcast Φ and receive neighbors' Φ_i
- $\Phi = \Phi + \sum_{i \in N} \Phi_i / d(k) - |N| \Phi / d(k)$



$\rho =$



MMC Epoch Example

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

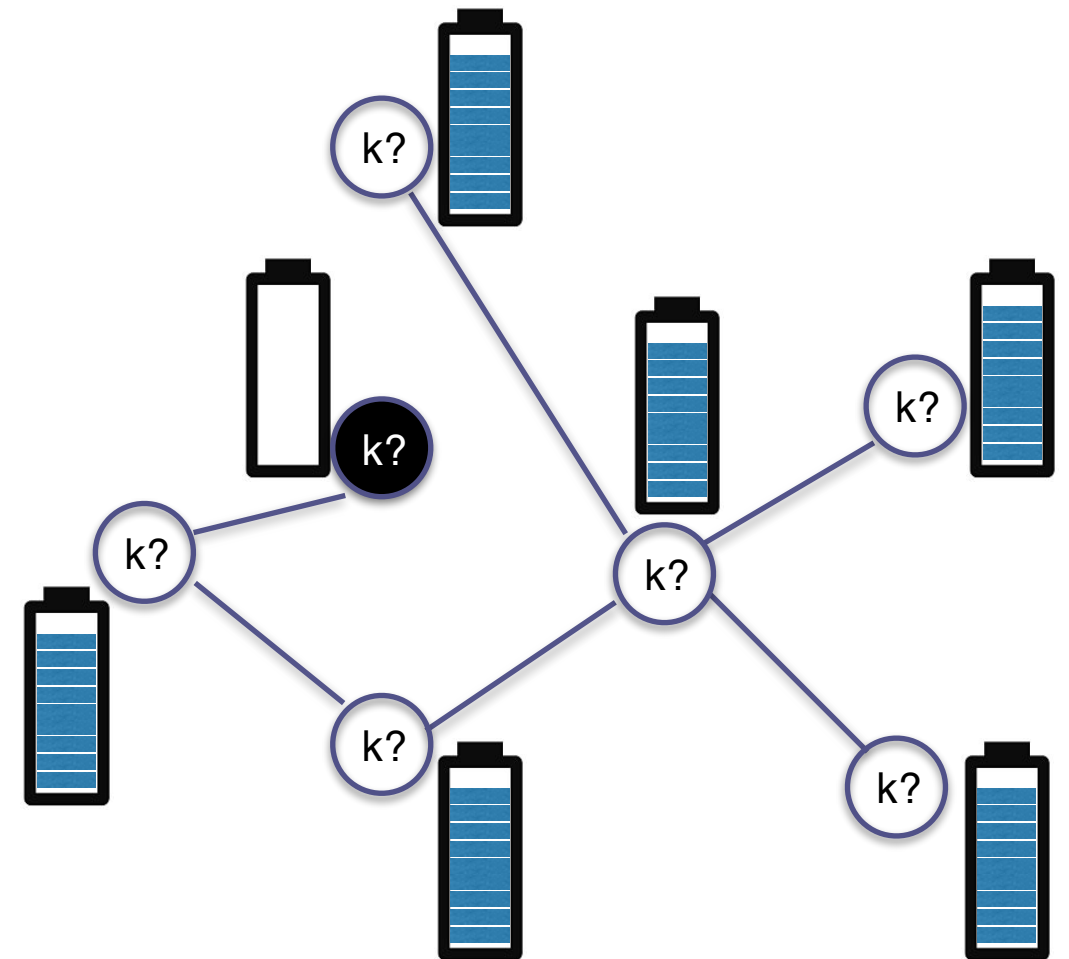
(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)

mass distribution:

- broadcast Φ and receive neighbors' Φ_i
- $\Phi = \Phi + \sum_{i \in N} \Phi_i / d(k) - |N| \Phi / d(k)$
- blacks “remove” their potential: $\rho = \rho + \Phi, \Phi = 0$



MMC Epoch Example

epochs:

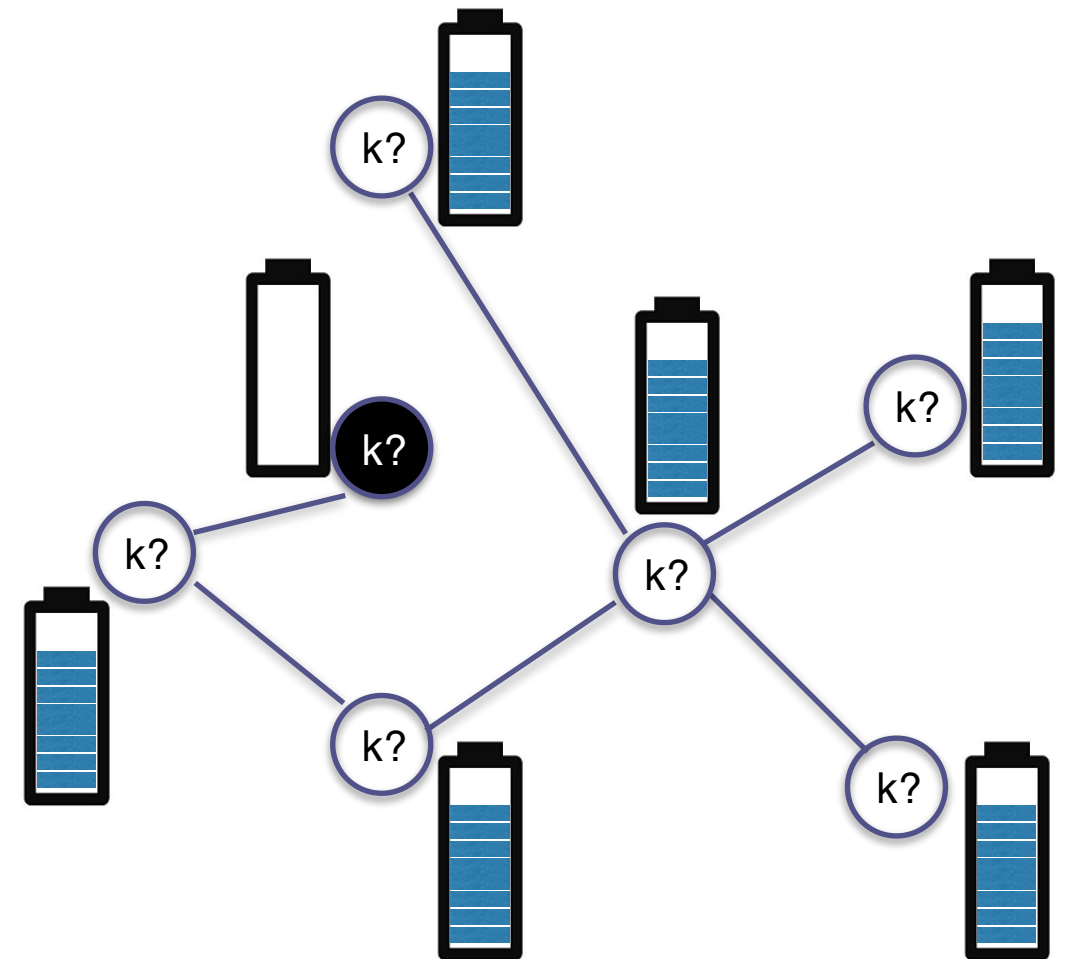
- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)



MMC Epoch Example

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

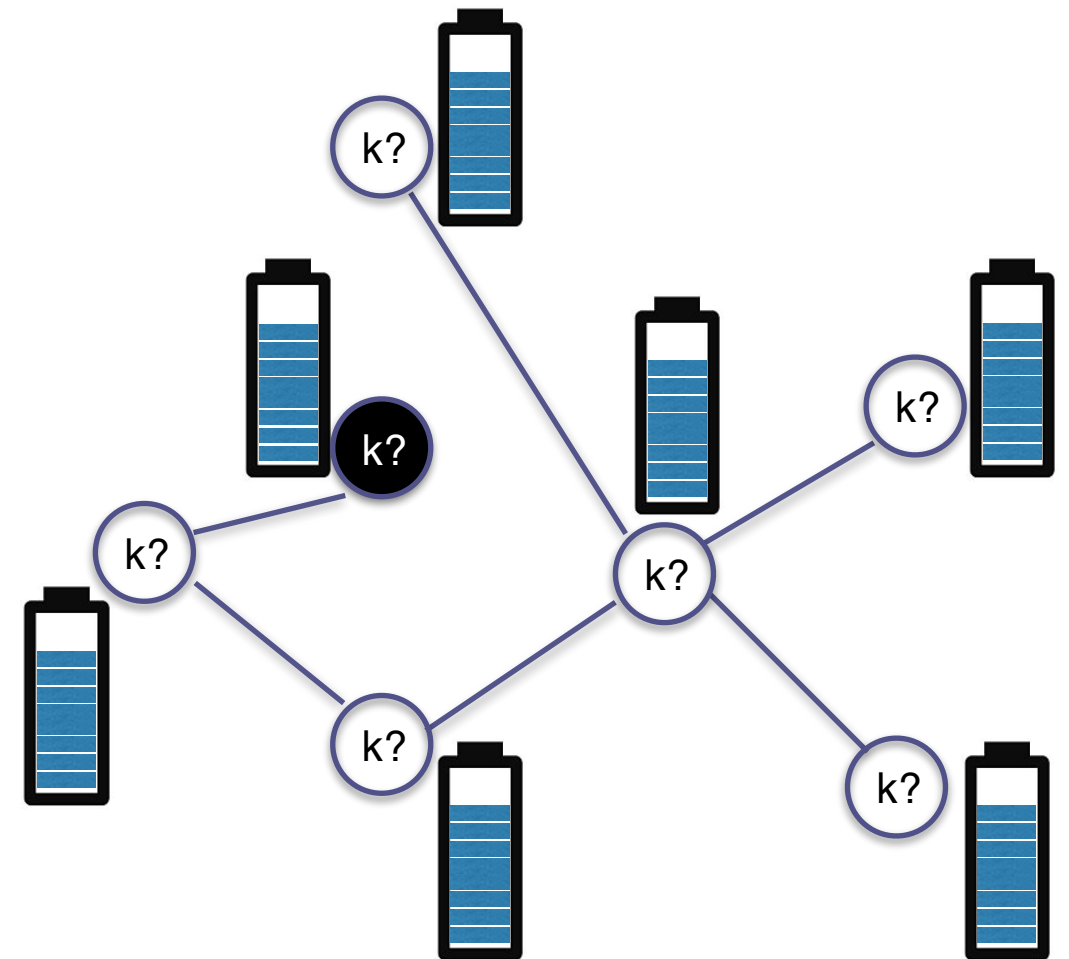
(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)

mass distribution:

- broadcast Φ and receive neighbors' Φ_i
- $\Phi = \Phi + \sum_{i \in N} \Phi_i / d(k) - |N| \Phi / d(k)$



MMC Epoch Example

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

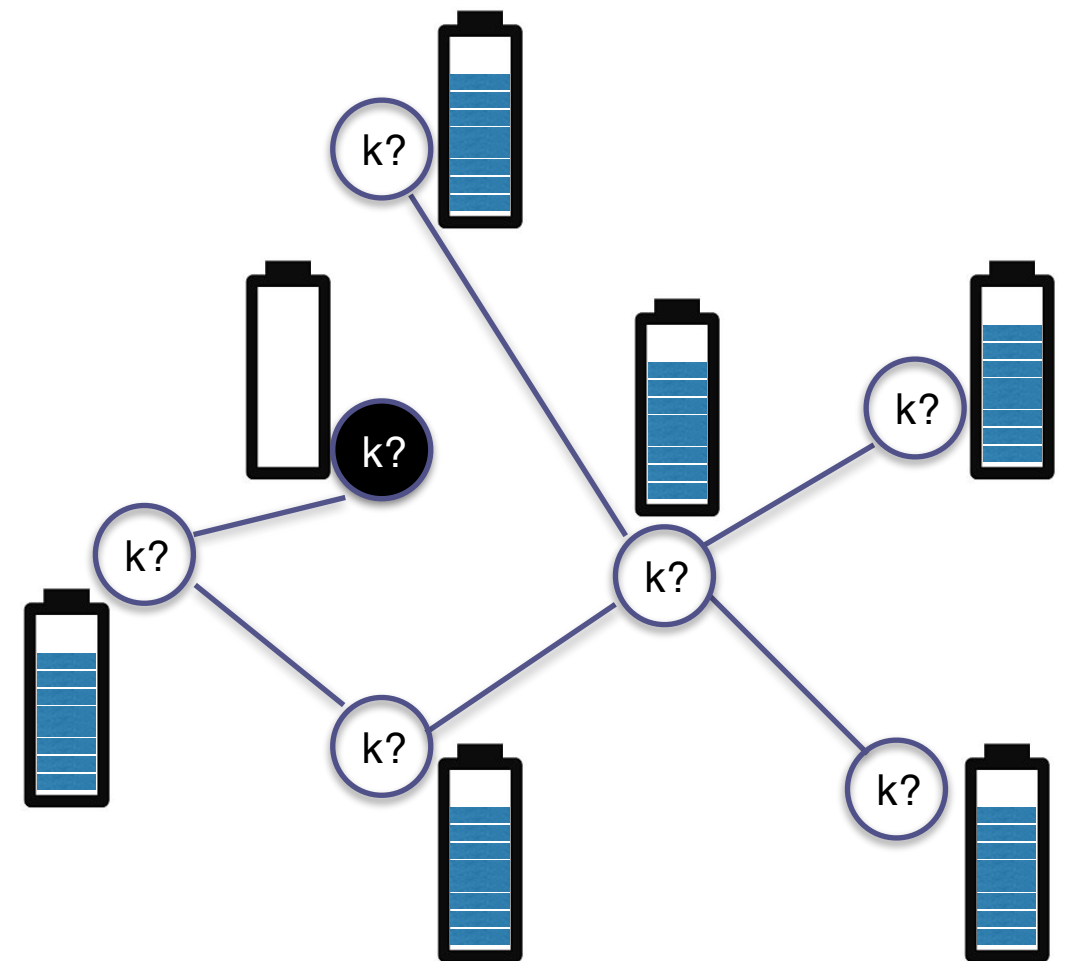
(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)

mass distribution:

- broadcast Φ and receive neighbors' Φ_i
- $\Phi = \Phi + \sum_{i \in N} \Phi_i / d(k) - |N| \Phi / d(k)$
- blacks “remove” their potential: $\rho = \rho + \Phi, \Phi = 0$



MMC Epoch Example

epochs:

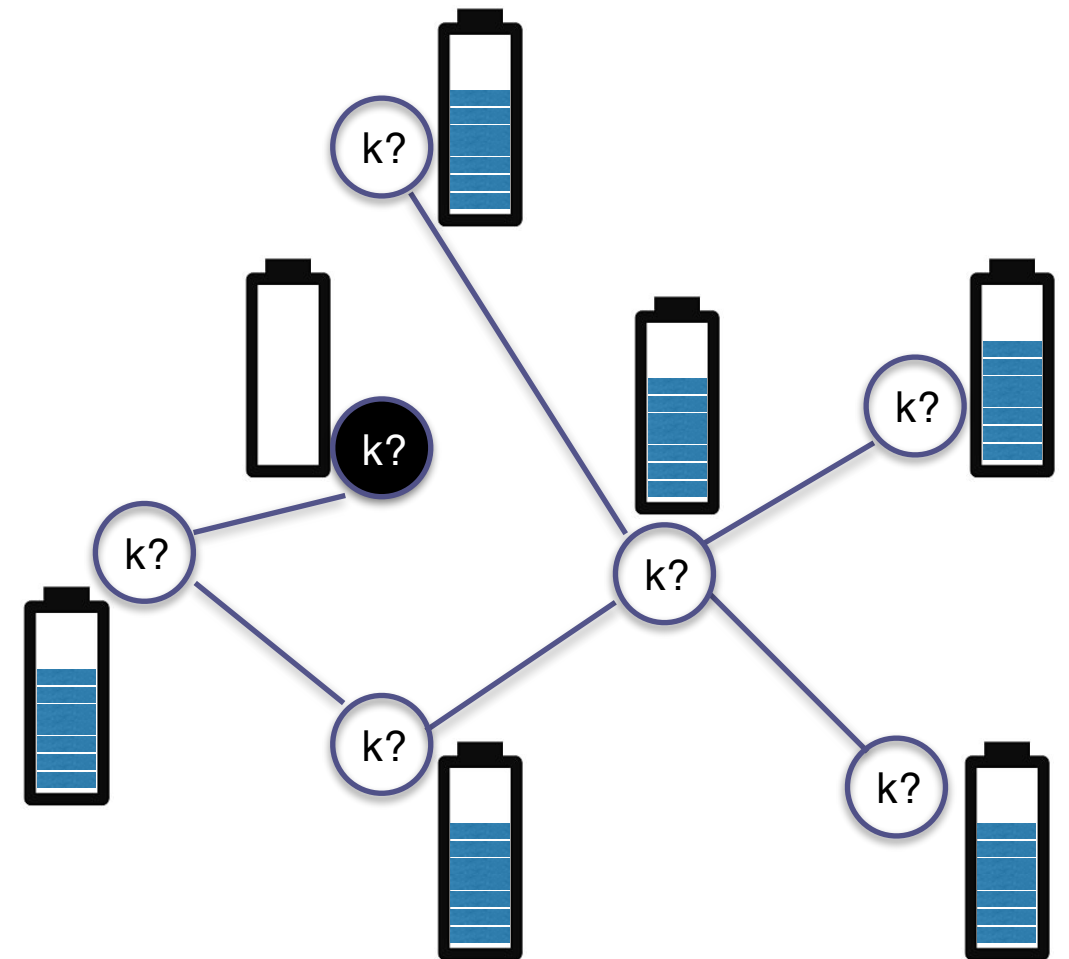
- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)



MMC Epoch Example

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

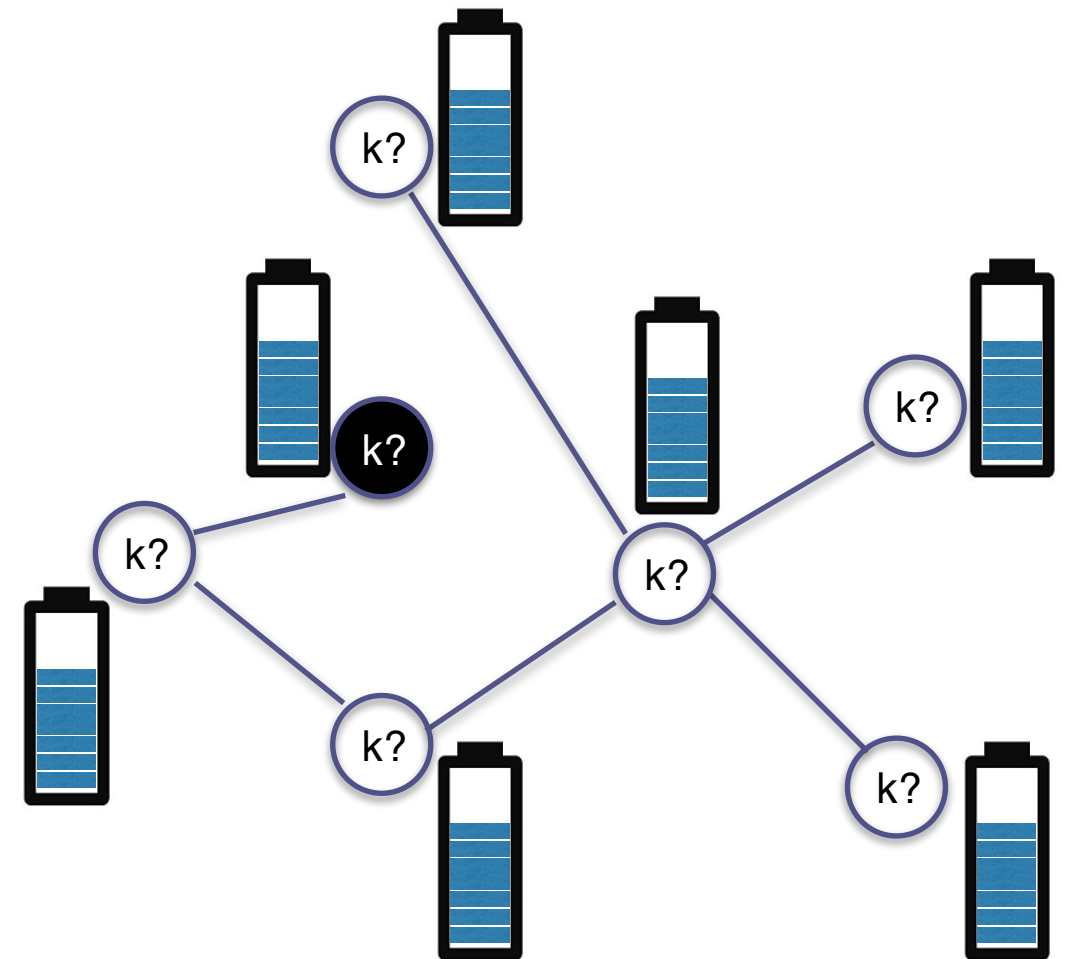
(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)

mass distribution:

- broadcast Φ and receive neighbors' Φ_i
- $\Phi = \Phi + \sum_{i \in N} \Phi_i / d(k) - |N| \Phi / d(k)$



MMC Epoch Example

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

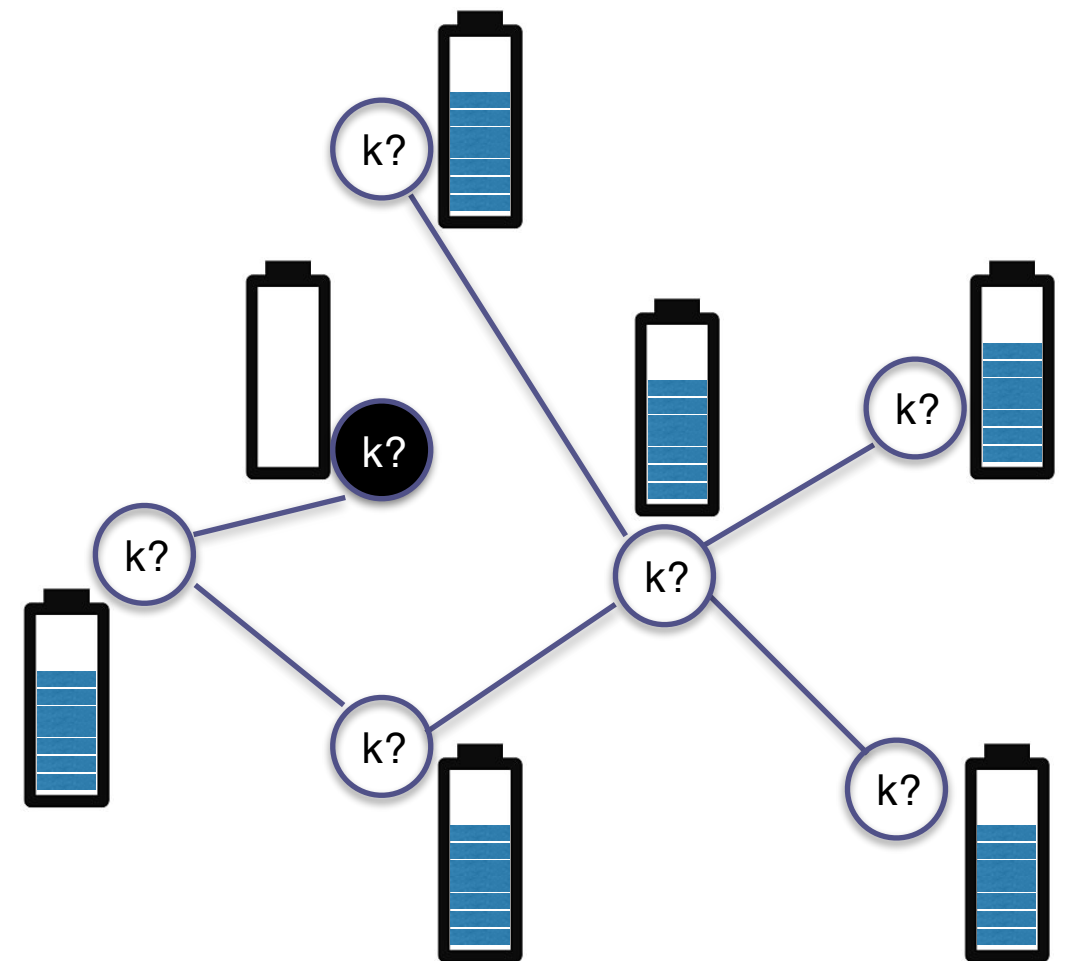
(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)

mass distribution:

- broadcast Φ and receive neighbors' Φ_i
- $\Phi = \Phi + \sum_{i \in N} \Phi_i / d(k) - |N| \Phi / d(k)$
- blacks “remove” their potential: $\rho = \rho + \Phi, \Phi = 0$



MMC Epoch Example

epochs:

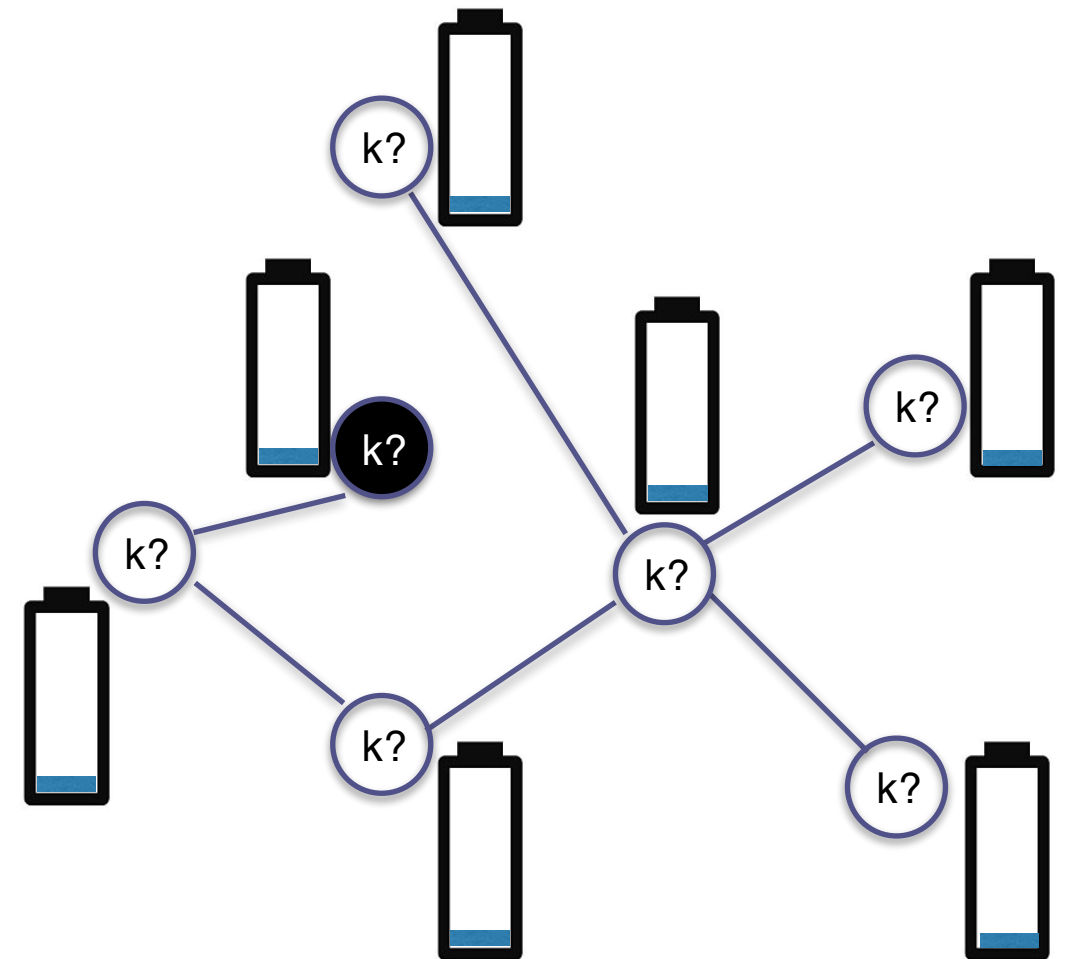
- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

(to let blacks remove “enough” potential ρ)

$r(k)$ rounds:

(to “average” the current potentials Φ)



$\rho =$



MMC Epoch Example

epochs:

- one for each estimate $k=l+1, 2(l+1), 4(l+1), \dots$
- initially, “potential” value: $\Phi_{\text{white}}=1, \Phi_{\text{black}}=0$

$p(k)$ phases:

(to let blacks remove “enough” potential ρ)

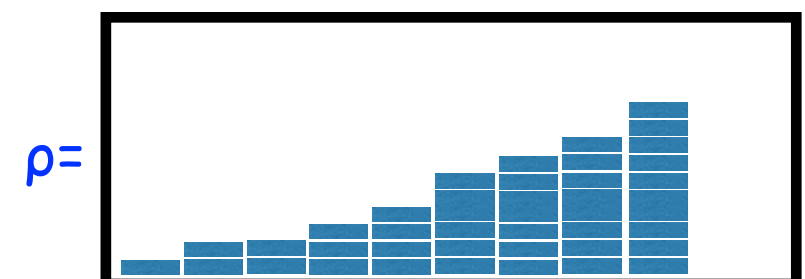
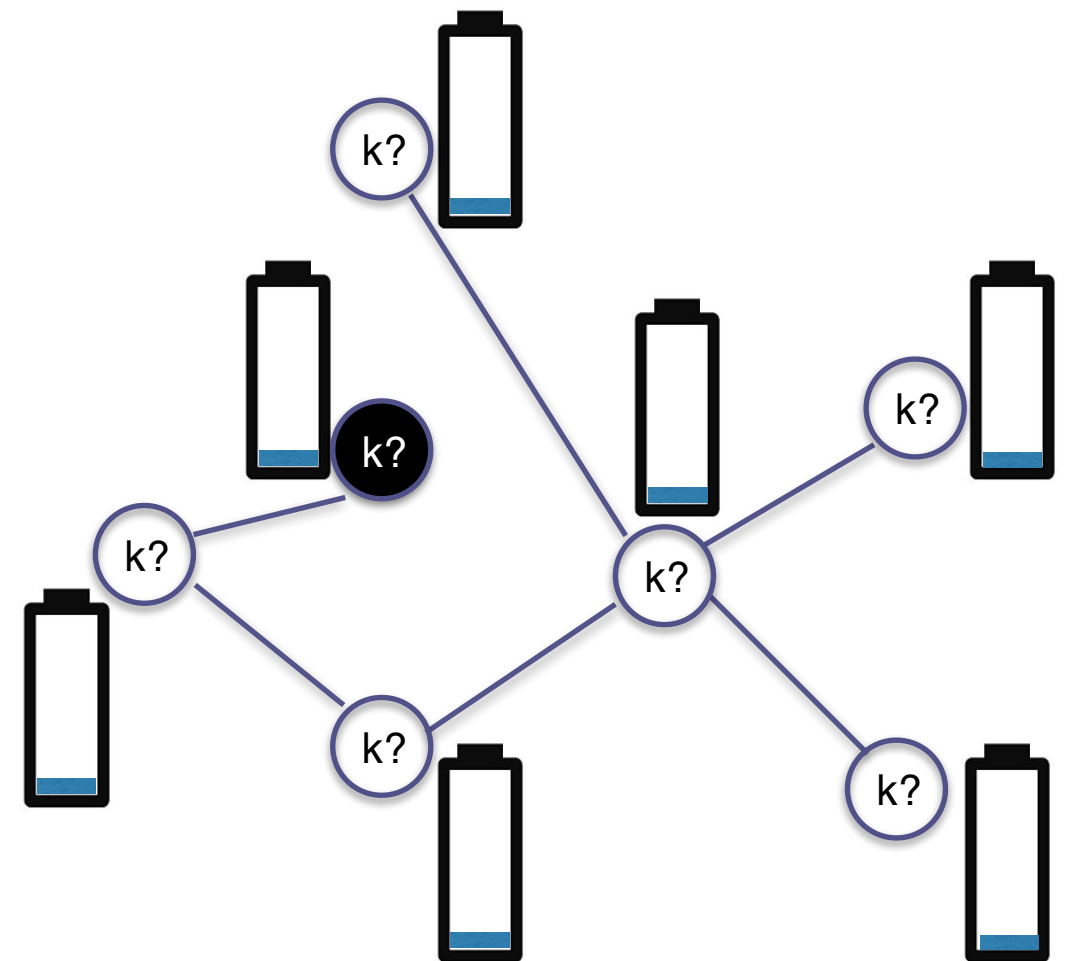
$r(k)$ rounds:

(to “average” the current potentials Φ)

mass distribution:

- broadcast Φ and receive neighbors' Φ_i
- $\Phi = \Phi + \sum_{i \in N} \Phi_i / d(k) - |N| \Phi / d(k)$
- blacks “remove” their potential: $\rho = \rho + \Phi, \Phi = 0$
- blacks decide according to ρ
- blacks notify if $k \geq n$
- try next k if needed

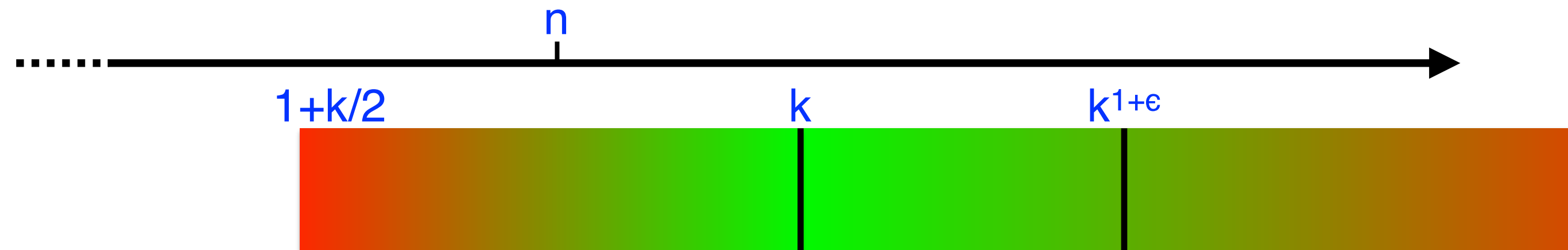
After $p(k)$ phases...



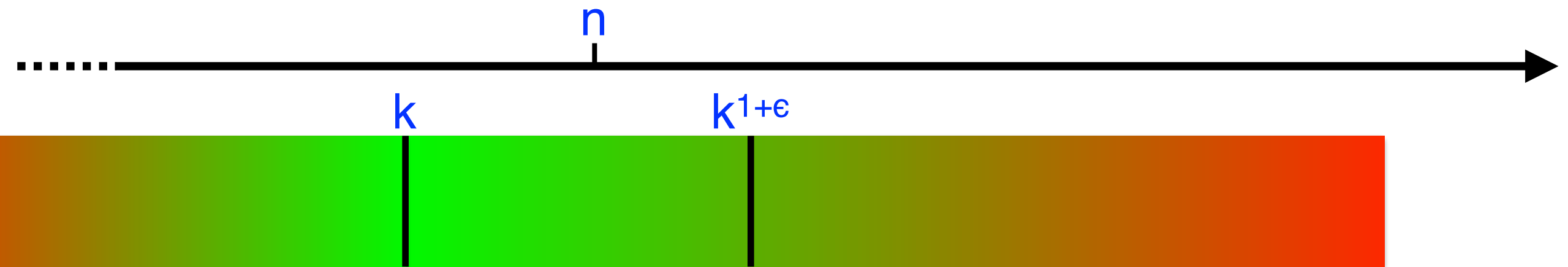
MMC Alarms



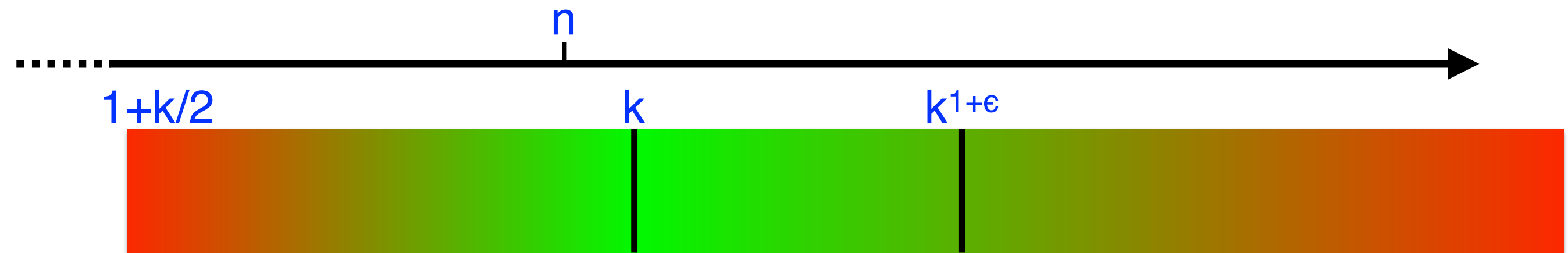
MMC Alarms



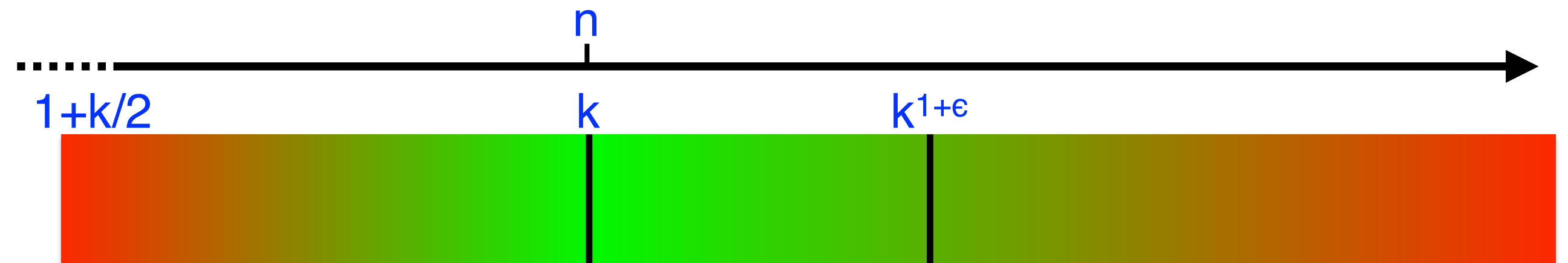
MMC Alarms



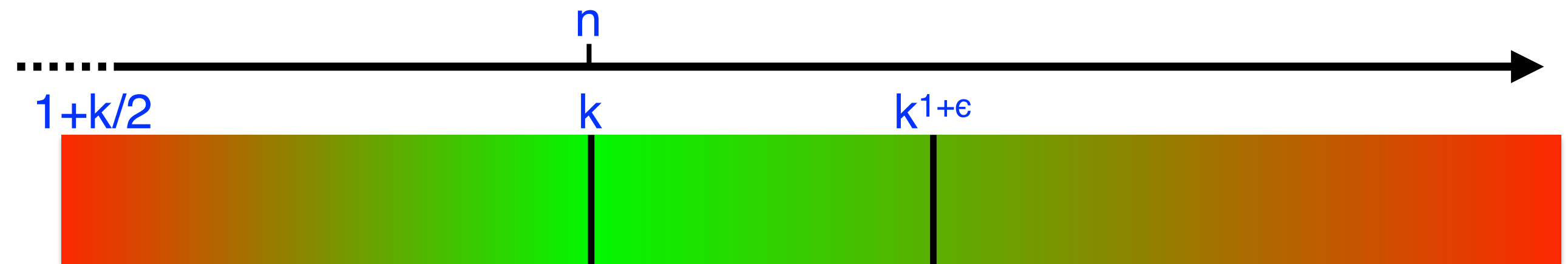
MMC Alarms



MMC Alarms



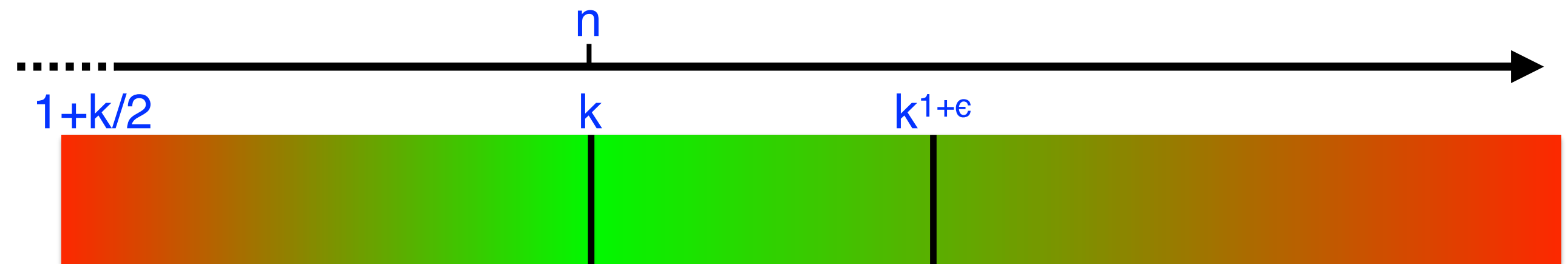
MMC Alarms



If n is “far” from k then not
“many” nodes have “low”
potential after phase 1,

so, blacks receive alarm from
nodes with “high” potential
“soon” after phase 1.

MMC Alarms



If n is “close” to k
from above
then blacks remove
“too much” potential.

If n is “far” from k then not
“many” nodes have “low”
potential after phase 1,

so, blacks receive alarm from
nodes with “high” potential
“soon” after phase 1.

MMC Alarms



If n below k then
blacks remove
“too little” potential.

If n is “close” to k
from above
then blacks remove
“too much” potential.

If n is “far” from k then not
“many” nodes have “low”
potential after phase 1,

so, blacks receive alarm from
nodes with “high” potential
“soon” after phase 1.

Randomized Counting

As we showed, if ℓ is unknown or zero, \exists executions that do not stop
 \Rightarrow we need black nodes, and we need to know how many,
we aim (stochastically) for $\ell=1$.

Randomized Counting

LLMC Key Ingredients:

- consider consecutive powers of 2 as values of K
- for each K
- each node chooses to be black with probability inverse of K
- run MMC for $\ell=1$ and $k \leq K$
- if $K \geq n$ and there is one black node \Rightarrow done

Randomized Counting

LLMC Key Ingredients:

- consider consecutive powers of 2 as values of K
- for each K
- each node chooses to be black with probability inverse of K
- run MMC for $\ell=1$ and $k \leq K$
- if $K \geq n$ and there is one black node \Rightarrow done

Challenges:

- How to detect that $K \geq n$?
- If $\ell=0$ no count, but what if $\ell>1$?

Randomized Counting

LLMC Key Ingredients:

- consider consecutive powers of 2 as values of K
- for each K
- each node chooses to be black with probability inverse of K
- run MMC for $\ell=1$ and $k \leq K$
- if $K \geq n$ and there is one black node \Rightarrow done

Two additional techniques:

- Run parallel threads:
 - if # threads with $\ell=0$ is large enough, $K \geq n$ is likely
- Take max count over threads:
 - count with $\ell>1$ is smaller than with $\ell=1$

LLMC

```
1: procedure
2:    $K \leftarrow \lceil \lceil 12/(\epsilon) \rceil \rceil$  //  $\lceil \lceil x \rceil \rceil$ : the smallest power of 2 bigger than  $x$ 
3:    $Count \leftarrow \emptyset$  // set of potentially "good" estimates computed in threads
4:    $EmptyThreads \leftarrow 0$  // number of threads with no black node detected
5:   while  $Count = \emptyset$  or  $EmptyThreads \leq f(K)/2$  do
6:      $Count \leftarrow \emptyset, EmptyThreads \leftarrow 0$ 
7:      $K \leftarrow 2K$ 
8:     Initiate  $f(K) = 64 \frac{\log(K/\epsilon)}{\log(e/(e-2))}$  parallel threads
        // parallel computation and messages sharing same resources/medium
9:     for each thread do
10:      for each node do
11:        | Select to be a black node with probability  $1/g(K)$ , where  $g(K) = K/2$ 
12:      end for
13:       $k \leftarrow MMC(K, 1)$  // refer to Figure 2
14:      if  $k > 0$  then
15:        |  $Count \leftarrow Count \cup \{k\}$ 
16:      end if
17:      if no black node detected then
18:        | Increase  $EmptyThreads$  by 1
19:      end if
20:    end for
21:  end while
22:  return  $\max(Count)$  // Output the maximum number in  $Count$  as the size  $n$ .
23: end procedure
```

LLMC

if K too small \Rightarrow
prob of black too high \Rightarrow
black nodes too big \Rightarrow
empty threads too small

```
1: procedure
2:    $K \leftarrow \lceil \lceil 12/(\epsilon) \rceil \rceil$  //  $\lceil \lceil x \rceil \rceil$ : the smallest power of 2 bigger than  $x$ 
3:    $Count \leftarrow \emptyset$  // set of potentially "good" estimates computed in threads
4:    $EmptyThreads \leftarrow 0$  // number of threads with no black node detected
5:   while  $Count = \emptyset$  or  $EmptyThreads \leq f(K)/2$  do
6:      $Count \leftarrow \emptyset, EmptyThreads \leftarrow 0$ 
7:      $K \leftarrow 2K$ 
8:     Initiate  $f(K) = 64 \frac{\log(K/\epsilon)}{\log(e/(e-2))}$  parallel threads
           // parallel computation and messages sharing same resources/medium
9:     for each thread do
10:      for each node do
11:        | Select to be a black node with probability  $1/g(K)$ , where  $g(K) = K/2$ 
12:      end for
13:       $k \leftarrow MMC(K, 1)$  // refer to Figure 2
14:      if  $k > 0$  then
15:        |  $Count \leftarrow Count \cup \{k\}$ 
16:      end if
17:      if no black node detected then
18:        | Increase  $EmptyThreads$  by 1
19:      end if
20:    end for
21:  end while
22:  return  $\max(Count)$  // Output the maximum number in  $Count$  as the size  $n$ .
23: end procedure
```

when K gets close to $n \Rightarrow$
 even if $\ell=1$, while $K < n \Rightarrow$
 it is $k \leq K < n \Rightarrow$
 no count

LLMC

if K too small \Rightarrow
 prob of black too high \Rightarrow
 # black nodes too big \Rightarrow
 # empty threads too small

```

1: procedure
2:    $K \leftarrow \lceil \lceil 12/\epsilon \rceil \rceil$  //  $\lceil \lceil x \rceil \rceil$ : the smallest power of 2 bigger than  $x$ 
3:    $Count \leftarrow \emptyset$  // set of potentially "good" estimates computed in threads
4:    $EmptyThreads \leftarrow 0$  // number of threads with no black node detected
5:   while  $Count = \emptyset$  or  $EmptyThreads \leq f(K)/2$  do
6:      $Count \leftarrow \emptyset, EmptyThreads \leftarrow 0$ 
7:      $K \leftarrow 2K$ 
8:     Initiate  $f(K) = 64 \frac{\log(K/\epsilon)}{\log(e/(e-2))}$  parallel threads
        // parallel computation and messages sharing same resources/medium
9:     for each thread do
10:      for each node do
11:        | Select to be a black node with probability  $1/g(K)$ , where  $g(K) = K/2$ 
12:      end for
13:       $k \leftarrow MMC(K, 1)$  // refer to Figure 2
14:      if  $k > 0$  then
15:        |  $Count \leftarrow Count \cup \{k\}$ 
16:      end if
17:      if no black node detected then
18:        | Increase  $EmptyThreads$  by 1
19:      end if
20:    end for
21:  end while
22:  return  $\max(Count)$  // Output the maximum number in  $Count$  as the size  $n$ .
23: end procedure
  
```

ICALP 2018 Open Questions

- Many distinguished nodes.
- Improve upper and/or lower bounds.
- Other computations in ADNs (beyond sum, avg, etc.).
- Asynchronous protocol.

ICALP 2018 Open Questions

- Many distinguished nodes. ✓
- Improve upper and/or lower bounds. ✓
- Other computations in ADNs (beyond sum, avg, etc.).
- Asynchronous protocol.

Thank you!

