## Counting with Population Protocols

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Introduced by Angluin, Aspnes, Diamadi, Fischer and Peralta in $2004{ }^{1}$

A model of distributed systems with minimal set of assumptions
(1) Finite-state agents / automata
(2) Agents have no identities
(3) Communication is occasionally possible between agents
$\rightarrow$ Uniformity All the agents execute the same code and this code is independent of the population size
$\rightarrow$ Anonymity: No room for agents to store a unique identifier

[^0]
## A motivating example: birds [Angluin04]

A flock of birds on which are strapped identical sensors When birds are close together, their sensors can interact and compute simple functions

- Decide if at least five birds have elevated body temperature
- Decide if at least five percent of the birds have elevated body temperature
- Compute the proportion of birds that have elevated temperature
- ...



## Agenda

- Motivation to study the population protocol model
- Model
- The counting problem
- Performance evaluation
- Conclusion and future works



## The population protocol model (informal)

- Collection of finitely many finite-state agents
- Agents are indistinguishable (identical program + no identity)
- Agents enter their initial state by applying an input function I to a finite set $X$ of input values
$I($ safe $)=q_{0}$
$1($ sick $)=q_{1}$



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All agents eventually converge to a correct common or distributed output value


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## The population protocol model (informal)

- A population configuration C specifies the state of each agent
- Fairness condition: enforces that any reachable configuration is eventually reached.
- A stable computation is a infinite fair sequence of configurations that converges to a correct common or distributed output value
- Convergence is a global property: agents generally do not know that convergence has been reached.
- With suitable stochastic assumptions, it is possible to determine the number of interactions until the output stabilizes


## A motivating example: birds [Angluin 04]

A flock of birds on which are strapped identical sensors
When birds are close together, their sensors can interact and compute simple functions

Decide if at least five birds have elevated body temperature ? Decide if at least five percent of the birds have elevated body temperature? Compute the proportion of birds that have elevated temperature ?


## Computational power of population protocols

- A predicate can be seen as a function that returns true or false
- Predicates can be written as $P\left(x_{1}, x_{2}, \ldots, x_{k}\right)$, where
- $k=$ number of possible initial states
- $x_{i}=$ number of agents starting in the $i$-th state
- Examples:
- the "count-to-5" bird protocol: $P\left(x_{1}, x_{2}\right)=$ true iff $x_{1} \geq 5$
- the "majority" bird protocol: $P\left(x_{1}, x_{2}\right)=$ true iff $x_{1} \geq x_{2}$
- the "parity" bird protocol: $P\left(x_{1}, x_{2}\right)=$ true iff $x_{1} \equiv 0(\bmod 2)$


## Computational power of population protocols

## Theorem (Angluin et al. 2007)

A predicate is computable by a population protocol if and only if it is in one of the following forms ${ }^{a}$ :

- $\sum_{i=1}^{k} c_{i} x_{i} \geq a$
- $\sum_{i=1}^{k} c_{i} x_{i} \equiv a(\bmod b)$,
where $a_{i}, b$ and $c_{i}^{\prime}$ s are integer constant
- And every boolean combination of these predicates

[^1]
# Our contribution: counting the exact percentage of sick and safe birds 

## Problem

Design of a population protocol that exactly determines the difference between the percentage of sick birds and the percentage of safe birds.

# Our algorithm to compute the exact difference between the percentage of \#red and 

## Algorithm

- Finite input alphabet $X=\{$ sick,safe $\}$
- Input function $I: I($ sick $)=m, I($ safe $)=-m$
- Finite set of states $Q=\{-m,-m+1, \ldots, m-1, m\}$
- Output function $O: O(q)=\lfloor 100 q / m+1 / 2\rfloor$
- Transition function

$$
f:\left(q_{1}, q_{2}\right) \rightarrow\left(\left\lfloor\left(q_{1}+q_{2}\right) / 2\right\rfloor,\left\lceil\left(q_{1}+q_{2}\right) / 2\right\rceil\right)
$$

## Problem

What is the difference between the percentage of sick and the percentage of safe birds?


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## Problem

What is the difference between the percentage of sick and the percentage of safe birds?

- $\sum_{i=1}^{5}=$ constant $=-20$.

As you will see this is an invariant of the protocol


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What is the difference between the percentage of sick and the percentage of safe birds?
$\lfloor 100 q / m+1 / 2\rfloor=\lfloor 100 \times(-4) / 20+1 / 2\rfloor=-20$


## Computing with randomized interactions

- The choice of the interacting agents is a non-deterministic choice
- Probabilistic assumptions: interactions are orchestrated by a uniform fair random scheduler:
- At each discrete time $t$, any two agents $i$ and $j$ are randomly chosen from a uniform distribution $p_{i, j}(t)$, with

$$
p_{i, j}(t)=\frac{1}{n(n-1)}
$$

## How long does it take for each agent to converge ?

Main ideas of the convergence proof.

- We show that at each step, the sum of the agents' state is constant. For every $t \geq 0$,

$$
\sum_{i=1}^{n} C_{t}^{(i)}=\sum_{i=1}^{n} C_{0}^{(i)}
$$

- Let $\ell=\frac{1}{n} \sum_{i=1}^{n} C_{t}^{(i)}$ and $\mathrm{L}=(\ell, \ldots, \ell)$, we show that

$$
E\left(\left\|C_{t}-L\right\|^{2}\right)=\left(1-\frac{1}{n-1}\right)^{t} E\left(\left\|C_{0}-L\right\|^{2}\right)+\frac{n}{4}
$$

## How long does it take for each agent to converge ?

Main ideas of the convergence proof (continued).

- Let $\kappa=(\#$ red $-\# b l u e) / 100$
- For all $\delta \in(0,1), m=\left\lceil\sqrt{2} n^{3 / 2} / \sqrt{\delta}\right\rceil$ and for all

$$
\begin{gathered}
t \geq(n-1)\left(5 \ln 2+3 \ln n-\ln \delta+\frac{2}{m-1}\right), \text { we show that } \\
P\left\{O\left(C_{t}^{(i)}\right)=\kappa, \text { for all } i=1, \ldots, n\right\} \geq 1-\delta
\end{gathered}
$$

- Thus the parallel convergence time to get $\kappa$ with any high probability is $O(\log n)$
- The knowledge of the population size $n$ allows us to directly derive \#red - \#blue


## Performance evaluation



Figure : Evolution of the configuration vector for a conserved advantage \#red - \#blue equal to $3 n / 5$. Settings: $n=2^{22}=4,19 \times 10^{6}$.

## Performance evaluation



Figure : Evolution of the configuration vector for a conserved advantage \#red - \#blue equal to 0 . Settings: $n=2^{22}=4,19 \times 10^{6}$.

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## Related works: Computation of the majority, i.e., whether

- [DV12]², [MNRS14] ${ }^{3}$ :
- Four-state protocol with an expected convergence parallel time in $O(\log n)$
- When \#red $\simeq$ \#blue, the convergence parallel time is infinite
${ }^{2}$ [DV12] M. Draief and M. Vojnovic, "Convergence speed of binary interval consensus", SIAM Journal on Control and Optimization, 50(3):1087:1097, 2012
${ }^{3}$ [MNRS14] G. Mertzios, et al., "Determining majority in networks with local interactions and very small local memory", ICALP, pp 871-882, 2014


## Related works: Computation of the majority, i.e., whether

- [AAE07] $]^{4},[P V V 09]^{5}$ :
- Three-state protocol with an expected convergence parallel time in $O(\log n)$
- Only when \#red - \#blue is in $O(\sqrt{\log n})$

[^2]
## Related works: Computation of the majority, i.e., whether

- [AGV15] ${ }^{6}$ :
- $(\log n)$-state protocol with an expected convergence parallel time in $O(\log n)$
- Whatever the difference between \#red and \#blue
- The authors show that a convergence in $O(\log n)$ interactions in expectation is a lower bound.
${ }^{6}$ [AFV15] D. Alistarh, R. Gelashvili, and M. Vojnovic, "Fast and exact majority in population protocols", Technical report MSR-TR-2015-13, Microsoft research, 2015
- The population protocol model: A simple but powerful enough model to describe distributed systems
- The exact computation power of these protocols has been determined
- We have proposed simple proofs to show that the convergence time for the counting problem is logarithmic in the size of the population


[^0]:    ${ }^{1}$ D. Angluin, J. Aspnes, Z. Diamadi, M. J. Fischer, R. Peralta, "Computation in networks of passively mobile finite-state sensors", PODC 2004: 290-299

[^1]:    ${ }^{\text {a }}$ D. Angluin, J. Aspnes, D. Eisenstat, E. Ruppert, " The computational power of population protocols", Distributed Computing 20(4): 279-304 (2007)

[^2]:    ${ }^{4}$ [AAE07] D. Angluin, J. Aspnes and D. Eisenstat, "A simple population protocol for fast Robust approximation majority", Distributed Computing, 20(4):279-305, 2007
    ${ }^{5}$ [PVV09] E. Perron, D. Vasudevan and M. Vojnovic, "Using three states for binary consensus on complete graphs", INFOCOM, pp 2527-3435, 2009

