Polylogarithmic-Time Leader Election in Population Protocols
Using Polylogarithmic States

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Abstract

Population protocols are networks of finite-state agents, interacting randomly, and updating their states using simple rules. Despite their extreme simplicity, these systems have been shown to cooperatively perform complex computational tasks, such as simulating register machines to compute standard arithmetic functions. The election of a unique leader agent is a key requirement in such computational constructions. Yet, the fastest currently known population protocol for electing a leader only has linear convergence time, and, it has recently been shown that no population protocol using a constant number of states per node may overcome this linear bound.

In this paper, we give the first population protocol for leader election with polylogarithmic convergence time, using polylogarithmic memory states per node. The protocol structure is quite simple: each node has an associated value, and is either a leader (still in contention) or a minion (following some leader). A leader keeps incrementing its value and “defeats” other leaders in one-to-one interactions, and will drop from contention and become a minion if it meets a leader with higher value. Importantly, a leader also drops out if it meets a minion with higher absolute value. While these rules are quite simple, the proof that this algorithm achieves polylogarithmic convergence time is non-trivial. In particular, the argument combines careful use of concentration inequalities with anti-concentration bounds, showing that the leaders’ values become spread apart as the execution progresses, which in turn implies that straggling leaders get quickly eliminated. We complement our analysis with empirical results, showing that our protocol converges extremely fast, even for large network sizes.

1 Introduction

Recently, there has been significant interest in modeling and analyzing interactions arising in biological or bio-chemical systems through an algorithmic lens. Several interesting computational models have been proposed for such networks, for example, the cellular automata model [Neu66], the stone-age distributed computing model [EW13], or the population model [AAD+06].

In particular, population protocols [AAD+06], which are the focus of this paper, consist of a set of \( n \) finite-state nodes interacting in pairs, where each interaction may update the states of both participants. The goal is to have all nodes converge on an output value, which represents the result of the computation, usually a predicate on the initial state of the nodes. The set of interactions occurring at each step is assumed to be decided by an adversarial scheduler, which is usually subject to some fairness conditions. The standard scheduler when computing convergence bounds is the probabilistic (uniform random) scheduler [AAE08, PVV09, DV12, MNRS14], which picks the next pair to interact uniformly at random in each step. We adopt this probabilistic scheduler model in this paper. (Some references refer to this model as the probabilistic population model.) The fundamental measure of convergence is parallel time, defined as the number of scheduler steps until convergence, divided by \( n \).
The class of predicates computable by population protocols is now well-understood \cite{AAD+06,AAE06,AAER07} to consist precisely of \textit{semilinear predicates}, i.e. predicates definable in first-order Presburger arithmetic. The first such construction was given in \cite{AAD+06}, and later improved in terms of convergence time in \cite{AAE06}. A parallel line of research studied the computability of deterministic functions in chemical reaction networks, which are also instances of population protocols \cite{CDS14}. All three constructions fundamentally rely on the election of a single initial \textit{leader} node, which co-ordinates phases of computation.

Reference \cite{AAD+06} gives a simple protocol for electing a leader from a uniform population, based on the natural idea of having leaders eliminate each other directly through symmetry breaking. Unfortunately, this strategy takes at least linear parallel time in the number of nodes \(n\): for instance, once this algorithm reaches two surviving leaders, it will require \(\Omega(n^2)\) additional interactions for these two leaders to meet. Reference \cite{AAE06} proposes a significantly more complex protocol, conjectured to be sub-linear, and whose convergence is only studied experimentally. These references posit the existence of a sublinear-time population protocol for leader election as a “pressing” open problem in the area. In fact, the existence of a poly-logarithmic leader election protocol would imply that \textit{any semilinear predicate} is computable in poly-logarithmic time by a \textit{uniform population} \cite{AAE06}.

Recently, Doty and Soloveichik \cite{DS15} showed that \(\Omega(n^2)\) expected interactions are \textit{necessary} for electing a leader in the classic probabilistic protocol model in which each node only has constant number of memory states (with respect to \(n\)). This negative result implies that computing semilinear predicates in leader-based frameworks is subject to the same lower bound. In turn, this motivates the question of whether faster computation is possible if the amount of memory per node is allowed to be a function of \(n\).

**Contribution:** In this paper, we solve this problem by proposing a new population protocol for leader election, which converges in \(O(\log^3 n)\) expected parallel time, using \(O(\log^3 n)\) memory states per node. Our protocol, called \textit{LM} for \textit{Leader-Minion}, roughly works as follows. Throughout the execution, each node is either a leader, meaning that it can still win, or a minion, following some leader. Each node state is associated to some \textit{absolute value}, which is a positive integer, and with a \textit{sign}, positive if the node is still in contention, and negative if the node has become a minion.

If two leaders meet, the one with the larger absolute value survives, and increments its value, while the other drops out, becoming a minion, and adopting the other node’s value, but with a negative sign. (If both leaders have the same value, they both increment it and continue.) If a leader meets a minion with smaller absolute value than its own, it increments its value, while the minion simply adopts the leader’s value, but keeps the negative sign. Conversely, if a leader meets a minion with larger absolute value than its own, then the leader drops out of contention, adopting the minion’s value, with negative sign. Finally, if two minions meet, they update their values to the maximum absolute value between them, but with a negative sign.

These rules ensure that, eventually, a single leader survives. While the protocol is relatively simple, the proof of poly-logarithmic time convergence is non-trivial. In particular, the efficiency of the algorithm hinges on the minion mechanism, which ensures that a leader with high absolute value can eliminate other contenders in the system, without having to directly interact with them.

Roughly, the argument is based on two technical insights. First, consider two leaders at a given time \(T\), whose (positive) values are at least \(\Theta(\log n)\) apart. Then, we show that, within \(O(\log n)\) parallel time from \(T\), the node holding the smaller value has become a minion, with constant probability. Intuitively, this holds since 1) this node will probably meet either the other leader or one of its minions within this time interval, and 2) it cannot increase its count fast enough to avoid defeat. For the second part of the argument, we show via anti-concentration that, after parallel time \(\Theta(\log^2 n)\) in the execution, the values corresponding to an arbitrary pair of nodes will be separated by at least \(\Omega(\log n)\).

We ensure that the values of nodes cannot grow beyond a certain threshold, and set the threshold in such a way that the total number of states is \(\Theta(\log^3 n)\). We show that with high probability the leader will be elected before the values of the nodes reach the threshold. In the other case, remaining leaders with
threshold values engage in a backup dynamics where minions are irrelevant and leaders defeat each other when they meet based on random binary indicators which are set using the randomness of the scheduler. This process is slower but correct and only happens with very low probability, allowing to conclude that the LM algorithm converges to a single leader within $O(\log^3 n)$ parallel time, both with high probability and in expectation, using $O(\log^3 n)$ states.

In population protocols, in every interaction, one node is said to be the initiator, the other is the responder, and the state update rules can use this distinction. In our protocol, this would allow a leader (the initiator in the interaction) to defeat another leader with the same value (the responder), and could also simplify the backup dynamics of our algorithm. However, our LM algorithm has the nice property that the state update rules can be made completely symmetric with regards to the initiator and responder roles.\footnote{For this reason, LM algorithm works for $n > 2$ nodes, because to elect a leader among two nodes it is necessary to rely on the initiator-responder role distinction.}

Summing up, we give the first poly-logarithmic time protocol for electing a leader from a uniform population. We note that $\Omega(n \log n)$ interactions seem intuitively necessary for leader election, as this number is required to allow each node to interact at least once. However, this idea fails to cover all possible reaction strategies if nodes are allowed to have arbitrarily many states.

We complement our analysis with empirical data, suggesting that the convergence time of our protocol is close to logarithmic, and that in fact the asymptotic constants are small, both in the convergence bound, and in the upper bound on the number of states the protocol employs.

**Related Work:** We restrict our attention to work in the population model. The framework of population protocols was formally introduced in reference [AAD\textsuperscript{+}06], to model interactions arising in biological, chemical, or sensor networks. It sparked research into its computational power [AAD\textsuperscript{+}06, AAE06, AAER07], and into the time complexity of fundamental tasks such as majority [AAE08, PVV09, DV12, MNRS14, AGV15], and leader election [AAD\textsuperscript{+}06, AAE06].\footnote{The best known upper bound for deterministic majority is of $O(\log n \log s + \log n/(\epsilon s))$ parallel time [AGV15], where $n$ is the number of nodes, $s$ is the number of states per node, and $\epsilon$ is the initial node difference between the two input states. The two problems are complementary, and no complexity-preserving transformations exist, to our knowledge.} References interested in computability consider an adversarial scheduler which is restricted to be fair, e.g., where each agent interacts with every other agent infinitely many times. For complexity bounds, the standard scheduler is uniform, scheduling each pair uniformly at random at each step, e.g., [AAE08, PVV09, DV12, MNRS14]. This model is also known as the probabilistic population model.

To the best of our knowledge, no population protocol for electing a leader with sub-linear convergence time was known before our work. References [AAD\textsuperscript{+}06, AAE06, CDS14] present leader-based frameworks for population computations, assuming the existence of such a node. The existence of such a sub-linear protocol is stated as an open problem in [AAD\textsuperscript{+}06, AAE06]. Reference [DH13] proposes a leader-less framework for population computation.

Recent work by Doty and Soloveichik [DS15] showed an $\Omega(n^2)$ lower bound on the number of interactions necessary for electing a leader in the classic probabilistic protocol model in which each node only has constant number of memory states with respect to the number of nodes $n$ [AAER07]. The proof of this result is quite complex, and makes use of the limitation that the number of states remains constant even as the number of nodes $n$ is taken to tend to infinity.

Thus, our algorithm can be interpreted as a complexity separation between population protocols which may only use constant memory per node, and protocols where the number of states is allowed to be a function of $n$.

A parallel line of research studied self-stabilizing population protocols, e.g., [AAFJ06, FJ06, SNY\textsuperscript{+}10], that is, protocols which can converge to a correct solution from an arbitrary initial state. It is known that stable leader election is impossible in such systems [AAFJ06]; references [FJ06, SNY\textsuperscript{+}10] circumvent this
impossibility by relaxing the problem semantics. Our algorithm is not affected by this result since it is not self-stabilizing.

2 Preliminaries

Population Protocols: We assume a population consisting of \(n\) agents, or nodes, each executing as a deterministic state machine with states from a finite set \(Q\), with a finite set of input symbols \(X \subseteq Q\), a finite set of output symbols \(Y\), a transition function \(\delta : Q \times Q \rightarrow Q \times Q\), and an output function \(\gamma : Q \rightarrow Y\). Initially, each agent starts with an input from the set \(X\), and proceeds to update its state following interactions with other agents, according to the transition function \(\delta\). For simplicity of exposition, we assume that agents have identifiers from the set \(V = \{1, 2, \ldots, n\}\), although these identifiers are not known to agents, and not used by the protocol.

The agents’ interactions proceed according to a directed interaction graph \(G\) without self-loops, whose edges indicate possible agent interactions. Usually, the graph \(G\) is considered to be the complete graph on \(n\) vertices, a convention we also adopt in this paper.

The execution proceeds in steps, or rounds, where in each step a new edge \((u, w)\) is chosen uniformly at random from the set of edges of \(G\). Each of the two chosen agents updates its state according to function \(\delta\).

Parallel Time: The above setup considers sequential interactions; however, in general, interactions between pairs of distinct agents are independent, and are usually considered as occurring in parallel. In particular, it is customary to define one unit of parallel time as \(n\) consecutive steps of the protocol.

The Leader Election Problem: In the leader election problem, all agents start in the same initial state \(A\), i.e. the only state in the input set \(X = \{A\}\). The output set is \(Y = \{\text{Win}, \text{Lose}\}\).

A population protocol solves leader election within \(\ell\) steps with probability \(1 - \phi\), if it holds with probability \(1 - \phi\) that for any configuration \(c : V \rightarrow Q\) reachable by the protocol after \(\geq \ell\) steps, there exists a unique agent \(i\) such that, (1) for the agent \(i\), \(\gamma(c(i)) = \text{Win}\), and, (2) for any agent \(j \neq i\), \(\gamma(c(j)) = \text{Lose}\).

3 The Leader Election Algorithm

In this section, we describe the LM leader election algorithm. The algorithm has an integer parameter \(m > 0\), which we set to \(\Theta(\log^3 n)\). Each state corresponds to an integer value from the set \([-m, -m + 1, \ldots, -2, -1, 1, 2, m - 1, m, m + 1]\). Respectively, there are \(2m + 1\) different states. We will refer to states and values interchangeably. All nodes start in the same state corresponding to value 1.

The algorithm, specified in Figure 1, consists of a set of simple deterministic update rules for the node state. In the pseudocode, the node states before an interaction are denoted by \(x\) and \(y\), while their new states are given by \(x'\) and \(y'\). All nodes start with value 1 and continue to interact according to these simple rules. We prove that all nodes except one will converge to negative values, and that convergence is fast with high probability. This solves the leader election problem since we can define \(\gamma\) as mapping only positive states to \(\text{Win}\) (a leader).

Since positive states translate to being a leader according to \(\gamma\), we call a node a contender if it has a positive value, and a minion otherwise. We present the algorithm in detail below. The state updates (i.e. the transition function \(\delta\)) of the LM algorithm are completely symmetric, that is, the new state \(x'\) depends on \(x\) and \(y\) (lines 2-4) exactly as \(y'\) depends on \(y\) and \(x\) (lines 5-7).

If a node is a contender and has absolute value not less than the absolute value of the interaction partner, then the node remains a contender and updates its value using the contend-priority function (lines 3 and 6). The new value will be one larger than the previous value except when the previous value was \(m + 1\), in which case the new value will be \(m\).

\[^{4}\text{Alternatively, } \gamma \text{ that maps only two states with values } m \text{ and } m + 1 \text{ to } \text{WIN} \text{ would also work, but we will work with positive "leader" states for the simplicity of presentation.}\]
Parameters:
m, an integer > 0, set to Θ(log^3 n)

State Space:
LeaderStates = {1, 2, ..., m − 1, m, m + 1},
MinionStates = {−1, −2, ..., −m + 1, −m}

Input: States of two nodes, x and y
Output: Updated states x' and y'

Auxiliary Procedures:
is-contender(x) = { true if x ∈ LeaderStates;
          false otherwise.
contend-priority(x, y) = { m if max( |x|, |y| ) = m + 1;
                          max(|x|, |y|) + 1 otherwise.
minion-priority(x, y) = { −m if max( |x|, |y| ) = m + 1;
                          −max(|x|, |y|) otherwise.

1 procedure update⟨x, y⟩
  2   if is-contender(x) and |x| ≥ |y| then
  3       x' ← contend-priority(x, y)
  4   else x' ← minion-priority(x, y)
  5   if is-contender(y) and |y| ≥ |x| then
  6       y' ← contend-priority(x, y)
  7   else y' ← minion-priority(x, y)

Figure 1: The state update rules for the LM algorithm.

If a node had a smaller absolute value than its interaction partner, or was a minion already, then the
node will be a minion after the interaction. It will set its value using the minion-priority function, to either
−max(|x|, |y|), or −m if the maximum was m + 1 (lines 4 and 7).

Values m + 1 and m are treated exactly the same way by minions (essentially corresponding to − m).
These values serve as a binary tie-breaker among the contenders that ever reach the value m, as will become
clear from the analysis.

4 Analysis
In this section, we provide a complete analysis of our leader election algorithm.

Notation: Throughout this proof, we denote the set of n nodes executing the protocol by V. We measure
execution time in discrete steps (rounds), where each step corresponds to an interaction. The configuration
at a given time t is a function c : V → Q, where c(v) is the state of the node v at time t. (We omit the
explicit time t when clear from the context.) We call a node contender when the value associated with its
state is positive, and a minion when the value is negative. As previously discussed, we also assume that
n > 2. Also, for presentation purposes, consider n to be a power of two.

We first prove that the algorithm never eliminates all contenders and that a configuration with a single
contender means that a leader is elected.

Lemma 4.1. There is always at least one contender in the system. After an execution reaches a configuration
with only a single node v being a contender, then from this point, v will have c(v) > 0 (mapped to WIN by
γ) in every reachable future configuration c, and there may never be another contender.

Proof. By the structure of the algorithm, a node starts as a contender and may become a minion during
an execution, but a minion may never become a contender again. Moreover, an absolute value associated
with the state of a minion node can only increase to an absolute value of an interaction partner. Finally, an
absolute value may never decrease except from $m + 1$ to $m$.

Let us assume for contradiction that an execution reaches a configuration where all nodes are minions. Consider this time point $T$ and let the maximum absolute value of the nodes be $u$. Because the minions cannot increase the maximum absolute value in the system, there must have been a contender with value $u$ during the execution before time $T$. In order for this contender to have become a minion by time $T$, it must have interacted with another node with an absolute value strictly larger than $u$. However, an absolute value of a node never decreases except from $m + 1$ to $m$, and despite existence of a larger absolute value than $u$ before time $T$, $u$ was the largest absolute value at time $T$, thus $u$ must be equal to $m$. But after such an interaction the second node that was in the state $m + 1$ remains a contender with value $m$. Now it must have interacted with yet another node of value $m + 1$ before time $T$ in order to itself become a minion, but then this another node is left as a contender with value $m$ and the same reasoning applies to it. Our proof follows by infinite descent.

Consequently, whenever there is a single contender in the system, it must have the largest absolute value. Otherwise, it could interact with a node with a larger absolute value and become a minion itself, contradicting the proof above that all nodes may never be minions. Because of this invariant, the only contender may never become a minion and the minions can never become contenders again. \qed

Now we turn our attention to the convergence speed (assuming $n > 2$) of the LM algorithm. Our goal is bound the number of rounds necessary to eliminate all except a single contender. In order for a contender to get eliminated, it must come across a larger value of another contender, the value possibly conducted through a chain of multiple minions via multiple interactions.

We first show by a rumor spreading argument that if the difference between the values of two contenders is large enough, then the contender with the smaller value will become a minion within the next $O(n \log n)$ rounds, with constant probability. Then we use anti-concentration bounds to establish that for any two fixed contenders, given that no absolute value in the system reaches $m$, after every $O(n \log^2 n)$ rounds the difference between their values is large enough with constant probability.

**Lemma 4.2.** Consider two contender nodes with values $u_1$ and $u_2$, where $u_1 - u_2 \geq 4\xi \log n$ at time $T$ for $\xi \geq 8$. Then, after $\xi n \log n$ rounds from $T$, the node that initially held the value $u_2$ will be a minion with probability at least $1/24$, independent of the history of previous interactions.

**Proof.** Call a node that has an absolute value of at least $u_1$ an up-to-date node, and out-of-date otherwise. Initially, at least one node is up-to-date. Before any round, if we have $x$ up-to-date nodes, the probability that an out-of-date node interacts with an up-to-date node in this round increasing the number of up-to-date nodes to $x + 1$, is $\frac{2x(x-n)}{n(n-1)}$. By a Coupon Collector argument, the expected number of rounds until every node is up-to-date is then $\sum_{x=1}^{n-1} \frac{n(n-1)}{2x} \leq \frac{n(n-1)}{2} \sum_{x=1}^{n-1} \left( \frac{1}{x} + \frac{1}{n-x} \right) \leq 2n \log n$.

Now by Markov’s inequality, the probability that not all nodes are up-to-date after $\xi n \log n$ communication rounds is at most $2/\xi$. Hence, expected number of nodes $Y$ that are up-to-date after $\xi n \log n$ communication rounds is at least $\frac{n(\xi-2)}{\xi}$. Let $q$ be the probability that the number of nodes after $\xi n \log n$ communication rounds is at least $\frac{n}{3} + 1$. Then we have $qn + (1 - q)(\frac{n}{3} + 1) \geq E[Y] \geq \frac{n(\xi-2)}{\xi}$, which implies that $q \geq \frac{1}{4}$ for $n > 2$ and $\xi \geq 8$.

Hence, with probability at least $1/4$, at least $n/3 + 1$ are nodes are up to date after $\xi n \log n$ rounds. By symmetry, the $n/3$ up-to-date nodes except the original node are uniformly random among the other $n - 1$ nodes. Therefore, any given node, in particular the node that had value $u_2$ at time $T$ has probability at least $1/4 \cdot 1/3 = 1/12$ to be up-to-date after $\xi n \log n$ rounds from $T$. Notice that when a node that was initially holding value $u_2$ becomes up-to-date and gets an absolute value of at least $u_1$ from an interaction, it must become a minion by the structure of the algorithm if its value before this interaction was still strictly smaller than $u_1$. Thus, we only need to show that the probability of selecting the node that initially had value $u_2$ at
least $4\xi \log n$ times (so that its value can reach $u_1$) during these $\xi n \log n$ rounds is at most $1/24$. The claim then follows by Union Bound.

In each round, the probability to select this node that initially held $u_2$ is $2/n$. Let us describe the number of times it is selected in $\xi n \log n$ rounds by considering a random variable $Z \sim \text{Bin}(\xi n \log n, 2/n)$. By Chernoff Bound, the probability being selected at least $4\xi \log n$ times in these rounds is at most:

$$\Pr[Z \geq 4\xi \log n] = \Pr[Z \geq (2\xi \log n)(1+1)] \leq \exp\left(-\frac{\xi}{3} \log n\right) \leq \frac{1}{n^{2\xi/3}} \leq \frac{1}{24},$$

finishing the proof. \hfill \Box

The next Lemma shows that, after $\Theta(n \log^2 n)$ rounds, the difference between the values of any two given contenders is high, with reasonable probability.

**Lemma 4.3.** Fix an arbitrary time $T$, and a constant $\xi \geq 1$. Consider any two contender nodes at time $T$, and time $T_1$ which is $32\xi^2 n \log^2 n$ rounds after $T$.

If no absolute value of any node reaches $m$ at any time until $T_1$, then, with probability at least $\frac{1}{24} - \frac{1}{n^{8\xi}}$, at time $T_1$, either at least one of the two nodes have become minions, or the absolute value of the difference of the two nodes’ values is at least $4\xi \log n$.

**Proof.** We will assume that no absolute value reaches $m$ at any point until time $T_1$ and that the two nodes are still contenders at $T_1$. We should now prove that the difference of values is large enough.

Consider the $32\xi^2 n \log^2 n$ rounds following time $T$. If a round involves an interaction with exactly one of the two fixed nodes we call it a spreading round. For each round, probability that it will be spreading is $\frac{4(n-2)}{n(n-1)}$ for $n > 2$ is at least $2/n$. So, we can describe the number of spreading rounds among the $32\xi^2 n \log^2 n$ rounds by considering a random variable $X \sim \text{Bin}(32\xi^2 n \log^2 n, 2/n)$. Then, by Chernoff Bound, the probability of having at most $32\xi^2 \log^2 n$ spreading rounds is at most

$$\Pr[X \leq 32\xi^2 \log^2 n] = \Pr\left[X \leq 64\xi^2 \log^2 n \left(1 - \frac{1}{2}\right)\right] \leq \exp\left(-\frac{64\xi^2 \log^2 n}{2^2 \cdot 2}\right) \leq 2^{-8\xi^2 \log^2 n} < \frac{1}{n^{8\xi}}.$$

Let us from now on focus on the high probability event that there are at least $32\xi^2 \log^2 n$ spreading rounds between times $T$ and $T_1$, and prove that the desired difference will be large enough with probability $\frac{1}{24}$. This implies the claim by Union Bound with the above event (note that for $n > 2$, $\frac{1}{n^{8\xi}} < \frac{1}{24}$ holds).

We assumed that both nodes remain contenders during the whole time, hence in each spreading round, a value of exactly one of them, with probability 1/2 each, increases by one. Without the loss of generality assume that at time $T$, the value of the first node was larger than or equal to the value of the second node. Let us now focus on the sum $S$ of $k$ independent uniformly distributed $\pm 1$ Bernoulli trials $x_i$ where $1 \leq i \leq k$, where each trial corresponds to a spreading round and outcome $+1$ means that the value of the first node increased, while $-1$ means that the value of the second node increased. In this terminology, we are done if we show that $\Pr[S \geq 4\xi \log n] \geq \frac{1}{24}$ for $k \geq 32\xi^2 \log^2 n$ trials.

However, we have that:

$$\Pr[S \geq 4\xi \log n] \geq \frac{\Pr[|S| \geq 4\xi \log n]}{2} = \frac{\Pr[|S|^2 \geq 16\xi^2 \log^2 n]}{2} \geq \frac{\Pr[|S|^2 \geq k/2]}{2} = \frac{\Pr[|S|^2 \geq \mathbb{E}[S^2]/2]}{2} \geq \frac{1}{2^2 \cdot 2 \cdot \mathbb{E}[S^2]} \geq \frac{1}{24},$$

where (4.1) follows from the symmetry of the sum with regards to the sign, that is, from $\Pr[S \geq 4\xi \log n] = \Pr[S < -4\xi \log n]$. For (4.2) we have used that $k \geq 32\xi^2 \log^2 n$ and $\mathbb{E}[S^2] = k$ (more about this below).
Finally, to get 4.3 we use Paley-Zygmund inequality and the fact that \( \mathbb{E}[S^4] = 3k(k-1) + k \leq 3k^2 \). Evaluating \( \mathbb{E}[S^2] \) and \( \mathbb{E}[S^4] \) is simple by using definition of \( S \) and the linearity of expectation. The expectation of each term then is either 0 or 1 and it suffices to count the number of terms with expectation 1, which are exactly the terms where each multiplier is raised to an even power. \( \square \)

Now we are ready to prove the convergence speed with high probability.

**Theorem 4.4.** There exists a constant \( \alpha \), such that for any constant \( \beta \geq 3 \) following holds: If we set \( m = \alpha \beta \log^3 n = \Theta(\log^3 n) \), the algorithm elects a leader (i.e. reaches a configuration with a single contender) in at most \( O(n \log^3 n) \) rounds (i.e. parallel time \( O(\log^3 n) \)) with probability at least \( 1 - 1/n^\beta \).

**Proof.** Let us fix \( \xi \geq 8 \) large enough such that
\[
\frac{1}{24} \cdot \left( \frac{1}{24} - \frac{1}{n^{8\xi}} \right) \geq p.
\]
for a constant \( 0 < p < 1 \). Let \( \beta \) be any constant \( \geq 3 \) and take \( \alpha = \frac{16}{p} \cdot (33\xi^2) \). We set \( m = \alpha \beta \log^3 n \) and consider the first \( \frac{\alpha \beta \log^3 n}{4} \) communication rounds of the algorithm execution. For any fixed node, the probability that it interacts in each round is \( 2/n \). Let us describe the number of times a given node interacts within the first \( \frac{\alpha \beta \log^3 n}{4} \) rounds by considering a random variable \( B \sim \text{Bin}(\frac{3\alpha \beta \log^3 n}{4}, 2/n) \). By Chernoff Bound, the probability being selected at least \( m \) times in these rounds is at most:
\[
\Pr [B \geq m] = \Pr \left[ B \geq \frac{\alpha \beta \log^3 n}{2} \cdot (1 + 1) \right] \\
\leq \exp \left( -\frac{\alpha \beta}{6} \log^3 n \right) \leq 2^{-\frac{\alpha \beta}{6} \log^3 n} \leq \frac{1}{n^{\alpha \beta/6}}.
\]
Taking Union Bound over all \( n \) nodes, with probability at least \( 1 - \frac{n}{n^{8\xi/4}} \), all nodes interact strictly less than \( m \) times during the first \( \frac{\alpha \beta \log^3 n}{4} \) rounds.

Let us from now on focus on the high probability event in above, which means that all absolute values are strictly less than \( m \) during the first \( \frac{\alpha \beta \log^3 n}{4} = \frac{4\beta}{p} (33\xi^2) n \log^3 n \) rounds. For a fixed pair of nodes, this allows us to apply Lemma 4.3 followed by Lemma 4.2 (with parameter \( \xi \))
\[
\frac{4\beta (33\xi^2) n \log^3 n}{p (32\xi^2 n \log^3 n + \xi n \log n)} \geq \frac{4\beta \log n}{p}
\]
times. Each time, first by Lemma 4.3, after \( 32\xi^2 n \log^3 n \) rounds with probability at least \( \frac{1}{24} - \frac{1}{n^{8\xi}} \) the nodes end up with values at least \( 4\xi \log n \) apart, in which case, after the next \( \xi n \log n \) rounds, by Lemma 4.2, one of the nodes becomes a minion with probability at least \( 1/24 \). Since Lemma 4.2 is independent from the interactions that precede it, by (4.4), each of the \( \frac{4\beta \log n}{p} \) times if both nodes are contenders, we get probability at least \( p \) that one of the nodes becomes a minion. Consider a random variable \( W \sim \text{Bin} \left( \frac{4\beta \log n}{p}, p \right) \). By Chernoff bound the probability that both nodes in a given pair are still contenders after the first \( \frac{\alpha \beta \log^3 n}{4} \) rounds is thus at most:
\[
\Pr [W \leq 0] = \Pr [W \leq 4\beta \log n (1 - 1)] \leq \exp \left( -\frac{4\beta \log n}{2} \right) \leq 2^{-2\beta \log n} \leq \frac{1}{n^{2\beta}}.
\]
By Union Bound over all \( \frac{n(n-1)}{2} < n^2 \) pairs the probability that for every pair of nodes, one of them is a minion after \( \frac{\alpha \beta \log^3 n}{4} \) communication rounds with probability at least \( 1 - \frac{n^2}{n^{8\xi}} - \frac{n}{n^{\alpha \beta/6}} \geq 1 - \frac{1}{n^\beta} \) for \( \beta \geq 3 \). A single contender means that leader is elected by Lemma 4.1. \( \square \)

Finally, we can prove the expected convergence bound.
Figure 2: The performance of the LM protocol. Both axes are logarithmic. The dots represent the results of individual experiments (100 for each network size), while the solid line represents the mean value for each network size.

Theorem 4.5. There is a setting of parameter $m$ of the algorithm such that $m = \Theta(\log^3 n)$, such that the algorithm elects the leader in expected $O(n \log^3 n)$ rounds of communication (i.e. parallel time $O(\log^3 n)$).

Proof. Let us prove that from any configuration, the algorithm elects a leader in expected $O(n \log^3 n)$ rounds. By Lemma 4.1, there is always a contender in the system and if there is only a single contender, then a leader is already elected. Now in a configuration with at least two contenders consider any two of them. If their values differ, then with probability at least $1/n^2$ these two contenders will interact in the next round and the one with the lower value will become a minion (after which it may never be a contender again). If the values are the same, then with probability at least $1/n$, one of these nodes will interact with one of the other nodes in the next round, leading to a configuration where the values of our two nodes differ, from where in the next round, independently, with probability at least $1/n^2$ these nodes meet and one of them again becomes a minion. Hence, unless a leader is already elected, in any case, in every two rounds, with probability at least $1/n^3$ the number of contenders decreases by 1.

Thus the expected number of rounds until the number of contenders decreases by 1 is at most $2n^3$. In any configuration there can be at most $n$ contenders, thus the expected number of rounds until reaching a configuration with only a single contender is at most $2(n - 1)n^3 \leq 2n^4$ from any configuration.

Now using Theorem 4.4 with $\beta = 4$ we get that with probability at least $1 - 1/n^4$ the algorithm converges after $O(n \log^3 n)$ rounds. Otherwise, with probability at most $1/n^4$ it ends up in some configuration from where it takes at most $2n^4$ expected rounds to elect a leader. The total expected number of rounds is therefore also $O(n \log^3 n) + O(1) = O(n \log^3 n)$, i.e. parallel time $O(\log^3 n)$. \qed

5 Experiments and Discussion

Empirical Data: We have also measured the convergence time of our protocol for different network sizes. (Figure 2 presents the results in the form of a log-log plot.) The protocol converges to a single leader quite fast, e.g., in less than 100 units of parallel time for a network of size $10^5$. This suggests that the constants hidden in the asymptotic analysis are small. The shape of the curve confirms the poly-logarithmic behavior of the protocol.

Discussion: We have given the first population protocol to solve leader election in poly-logarithmic time, using a poly-logarithmic number of states per node. Together with the results of [AAE06], the existence of

\footnote{This is always true, even when the new value is not larger, for instance when the values were equal to $m + 1$, the new value of one of the nodes will be $m \neq m + 1$.}
our protocol implies that population protocols can compute any semi-linear predicate on their input in time $O(n \log^5 n)$, with high probability, as long as memory per node is poly-logarithmic.

Our result opens several avenues for future research. The first concerns lower bounds. We conjecture that the lower bound for leader election in population protocols is $\Omega(\log n)$, irrespective of the number of states used by the protocol. Further, empirical data suggests that the analysis of our algorithm can be further tightened, cutting off logarithmic factors. It would also be interesting to prove a tight a trade-off between the amount of memory available per node and the running time of the protocol.

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