# Interpolating Arithmetic Read-Once Formulas in Parallel 

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#### Abstract

A formula is read-once if each variable appears at most once in it. An arithmetic read-once formula is one in which the operations are addition, subtraction, multiplication, and division (and constants are allowed). We present a randomized (Las Vegas) parallel algorithm for the exact interpolation of arithmetic read-once formulas over sufficiently large fields. More specifically, for $n$-variable read-once formulas, and fields of size at least $3\left(n^{2}+3 n-2\right)$, our algorithm runs in $O\left(\log ^{2} n\right)$ parallel steps using $O\left(n^{4}\right)$ processors (where the field operations are charged unit cost). This complements other results which imply that other classes of read-once formulas cannot be interpolated-or even learned with membership and equivalence queries-in poly-logarithmic-time with polynomially many processors (even though they can be learned sequentially in polynomial-time). These classes include boolean read-once formulas and arithmetic read-once formulas over fields of size $o(n / \log n)$ (for $n$ variable read-once formulas).


## 1 Introduction

The problem of interpolating a formula (from some class $C$ ) is the problem of exactly identifying the formula from queries to the assignment (membership) oracle. The interpolation algorithm queries the oracle with an assignment $a$ and the oracle returns the value of the function at $a$.

There are a number of classes of arithmetic formulas that can be interpolated sequentially in polynomial-time as well as in parallel in poly-logarithmic-time (with polynomially many processors). These include sparse polynomials and sparse rational functions ([BT88,BT90,GKS90,GrKS88,RB89,M91]).

[^0]A formula over a variable set $V$ is read-once if each variable appears at most once in it. An arithmetic read-once formula over a field $\mathcal{K}$ is a read-once formula over the basic operations of the field $\mathcal{K}$ : addition, subtraction, multiplication, division, and constants are also permited in the formula.

Bshouty, Hancock and Hellerstein [BHH92] present a randomized sequential polynomial-time algorithm for interpolating arithmetic read-once formulas (AROFs) over sufficiently large fields. Moreover, they show that, for arbitrarily-sized fields, arithmetic read-once formulas can be learned using equivalence queries in addition to membership queries.

The question of whether arithmetic read-once formulas can be interpolated (or learned) quickly in parallel depends on the size of the underlying field. It is shown in [BHH92] that for arithmetic read-once formulas over fields with $o(n / \log n)$ elements there is no poly-logarithmic-time algorithm that uses polynomially many processors (for interpolating as well as learning). Also, a similar negative result holds for boolean read-once formulas.

We present a (Las Vegas) parallel algorithm for the exact interpolation of arithmetic read-once formulas over sufficiently large fields. For fields of size at least $3\left(n^{2}+3 n \Leftrightarrow 2\right)$, the algorithm runs in $O\left(\log ^{2} n\right)$ parallel steps using $O\left(n^{4}\right)$ processors (where the field operations are charged unit cost).

If the "obvious" parallizations are made to the interpolating algorithm in [BHH92] (i.e., parallelizations of independent parts of the computation) one obtains a parallel running time that is $\Theta(d)$, where $d$ is the depth of the target formula. Since, in general, $d$ can be as large as $\Theta(n)$, this does not result in significant speedup. Our parallel algorithm uses some techniques from the sequential algorithm of [BHH92] as well as some new techniques that enable nonlocal features of the AROF to be determined in poly-logarithmic-time.

The parallel algorithm can be implemented on an oracle EREW PRAM that initially selects some random input values (uniformly and independently distributed) and then performs $O\left(n^{3}\right)$ membership queries (via its oracle).

## 2 Identification with queries

The learning criterion we consider is exact identification. There is a formula $f$ called the target formula, which is a member of a class of formulas $C$ defined over the variable set $V$. The goal of the learning algorithm is to halt and output a formula $h$ from $C$ that is equivalent to $f$.

In a membership query, the learning algorithm supplies values $\left(x_{1}^{(0)}, \ldots, x_{n}^{(0)}\right)$ for the variables in $V$ As input to a membership oracle, and receives in return the value of $f\left(x_{1}^{(0)}, \ldots, x_{n}^{(0)}\right)$. Note that if $f^{\prime}$ is a projection of $f$, it is possible to simulate a membership oracle for $f^{\prime}$ using a membership oracle for $f$.

We say that the class $C$ is learnable in polynomial time if there is an algorithm that uses the membership oracle and interpolates any $f \in C$ in polynomial time in the number of variables $n$ and the size of $f$. We say that $C$ is efficiently
learnable in parallel if there is a parallel algorithm that uses the membership oracle and interpolates any $f \in C$ in polylogarithmic time with polynomial number of processors. In the parallel computation $p$ processors can ask $p$ membership queries in one step.

## 3 Preliminaries

A formula is a rooted tree whose leaves are labeled with variables or constants from some domain, and whose internal nodes, or gates, are labeled with elements from a set of basis functions over that domain. A read-once formula is a formula for which no variable appears on two different leaves. An arithmetic read-once formula over a field $\mathcal{K}$ is a read-once formula over the basis of addition, subtraction, multiplication, and division of field elements, whose leaves are labeled with variables or constants from $\mathcal{K}$.

In [BHH92] it is shown that a modified basis can be used to represent any arithmetic read-once formula. Let $\mathcal{K}$ be an arbitrary field. The modified basis for arithmetic read-once formulas over $\mathcal{K}$ includes only two non-unary functions, addition $(+)$ and multiplication $(\times)$. The unary functions in the basis are $(a x+$ $b) /(c x+d)$ for every $a, b, c, d \in \mathcal{K}$ such that $a d \Leftrightarrow b c \neq 0$. This requirement is to prevent $a x+b$ and $c x+d$ being identically 0 or differing by just a constant factor. We can also assume that non-constant formulas over this modified basis do not contain constants in their leaves. We represent such a unary function as $f_{A}$, where

$$
A=\left(\begin{array}{ll}
a & b \\
c & d
\end{array}\right)
$$

The restriction on $a, b, c$, and $d$ is equivalent to saying the determinant of $A$ (denoted $\operatorname{det}(A)$ ) is non-zero.

The value of a read-once formula on an assignment to its variables is determined by evaluating the formula bottom up. This raises the issue of division by zero. In [BHH92] this problem is handled by defining basis functions over the extended domain $\mathcal{K} \cup\{\infty$, ERROR $\}$, where $\infty$ represents $1 / 0$ and ERROR represents $0 / 0$. For the special values we define our basis function as follows (assume $x \in \mathcal{K} \Leftrightarrow\{0\}$, $y \in \mathcal{K} \cup\{\infty, \operatorname{ERROR}\}$, and $A$ is as above).

$$
\begin{gathered}
y+\text { ERROR }=y \times \text { ERROR }=f_{A}(\text { ERROR })=\text { ERROR } \\
x+\infty=x \times \infty=\infty \\
0+\infty=\infty \times \infty=\infty \\
f_{A}(\infty)=\left\{\begin{array}{ll}
0 \times \infty=\infty+\infty=\text { ERROR } \\
\frac{a}{c} & c \neq 0 \\
\infty & c=0
\end{array} \text { and } f_{A}\left(\frac{-d}{c}\right)=\infty \text { if } c \neq 0\right.
\end{gathered}
$$

It is shown in [BHH92] that these definitions are designed so that the output of the read-once formula is the same as it would be if the formula were first expanded
and simplified to be in the form $p\left(x_{1}, \ldots, x_{n}\right) / q\left(x_{1}, \ldots, x_{n}\right)$ for some polynomials $p$ and $q$ where $\operatorname{gcd}(p, q)=1$, and then evaluated.

We say that a formula $f$ is defined on the variable set $V$ if all variables appearing in $f$ are members of $V$. Let $V=\left\{x_{1}, \ldots, x_{n}\right\}$. We say a formula $f$ depends on variable $x_{i}$ if there are values $x_{1}^{(0)}, x_{2}^{(0)}, \ldots, x_{n}^{(0)}$ and $x_{i}^{(1)}$ in $\mathcal{K}$ for which

$$
f\left(x_{1}^{(0)}, x_{2}^{(0)}, \ldots, x_{n}^{(0)}\right) \neq f\left(x_{1}^{(0)}, \ldots, x_{i-1}^{(0)}, x_{i}^{(1)}, x_{i+1}^{(0)}, \ldots, x_{n}^{(0)}\right)
$$

and for which both those values of $f$ are not ERROR. We call such an input vector $v=\left(x_{1}^{(0)}, \ldots, x_{n}^{(0)}\right)$ a justifying assignment for $x_{i}$.

Between any two gates or leaves $\alpha$ and $\beta$ in an AROF, the relationships ancestor, descendant, parent, and child refer to their relative position in the rooted tree. Let $\alpha \leq \beta$ denote that $\alpha$ is a descendant of $\beta$ (or, equivalently, that $\beta$ is an ancestor of $\alpha$ ). Let $\alpha<\beta$ denote that $\alpha$ is a proper descendant of $\beta$ (i.e., $\alpha \leq \beta$ but $\alpha \neq \beta$ ). For any pair of variables $x_{i}$ and $x_{j}$ that appear in a read-once formula, there is a unique node farthest from the root that is an ancestor of both $x_{i}$ and $x_{j}$, called their lowest common ancestor, which we write as lca $\left(x_{i}, x_{j}\right)$. We shall refer to the type of lca $\left(x_{i}, x_{j}\right)$ to mean the basis function computed at that gate. We say that a set $W$ of variables has a common lca if there is a single node that is the lca of every pair of variables in $W$.

We define the skeleton of a formula $f$ to be the tree obtained by deleting any unary gates in $f$ (i.e. the skeleton describes the parenthesization of an expression with the binary operations, but not the actual unary operations or embedded constants).

We now list a basic property of unary functions $f_{A}$ that is proved in [BHH92].

## Property 1

1. The function $f_{A}$ is a bijection from $\mathcal{K} \cup\{\infty\}$ to $\mathcal{K} \cup\{\infty\}$ if and only if $\operatorname{det}(A) \neq 0$. Otherwise, $f_{A}$ is either a constant value from $\mathcal{K} \cup\{\infty, E R R O R\}$ or else is a constant value from $\mathcal{K} \cup\{\infty\}$, except on one input value on which it is ERROR.
2. The functions $f_{A}$ and $f_{\lambda A}$ are equivalent for any $\lambda \neq 0$.
3. Given any three distinct points $p_{1}=\left(x_{1}, y_{1}\right), p_{2}=\left(x_{2}, y_{2}\right)$ and $p_{3}=\left(x_{3}, y_{3}\right)$,
(a) If $p_{1}, p_{2}, p_{3}$ are on a line then there exists a unique function $f_{A}$ with $f_{A}(x)=a x+b$ that satisfies $f_{A}\left(x_{1}\right)=y_{1}, f_{A}\left(x_{2}\right)=y_{2}$, and $f_{A}\left(x_{3}\right)=y_{3}$.
(b) If $p_{1}, p_{2}, p_{3}$ are not on a line then there exists a unique function $f_{A}$ with $\operatorname{det}(A) \neq 0$ that satisfies $f_{A}\left(x_{1}\right)=y_{1}, f_{A}\left(x_{2}\right)=y_{2}$, and $f_{A}\left(x_{3}\right)=y_{3}$.
4. If functions $f_{A}$ and $f_{B}$ are equivalent and $\operatorname{det}(A)$, $\operatorname{det}(B) \neq 0$, then there is a constant $\lambda$ for which $\lambda A=B$.
5. The functions $\left(f_{A} \circ f_{B}\right)$ and $f_{A B}$ are equivalent.
6. If $\operatorname{det}(A) \neq 0$, functions $f_{A}^{-1}$ and $f_{A^{-1}}$ are equivalent.
7. $f_{A}(\lambda x)=f_{A}\left(\begin{array}{cc}\lambda & 0 \\ 0 & 1\end{array}\right)(x)$ and $f_{A}(\lambda+x)=f_{A}\left(\begin{array}{cc}1 & \lambda \\ 0 & 1\end{array}\right)(x)$.

$$
\lambda f_{A}(x)=f_{\left(\begin{array}{cc}
\lambda & 0 \\
0 & 1
\end{array}\right) A_{A}}(x) \text { and } \lambda+f_{A}(x)=f_{\left(\begin{array}{ll}
1 & \lambda \\
0 & 1
\end{array}\right) A}(x) .
$$

## 4 Collapsability of Operations

Whenever two non-unary gates of the same type in an AROF are separated by only a unary gate it may be possible to collapse them together to a single non-unary gate of the same type with higher arity. For $\star \in\{+, \times\}$, a unary operation $f_{A}$ is called $\star$-collapsible if

$$
f_{A}(x \star y) \star z \equiv f_{B}(x) \star f_{C}(y) \star z,
$$

for some unary operations $f_{B}$ and $f_{C}$. Intuitively, the above property means that if the $f_{A}$ gate occurs between two non-unary $\star$ gates then the two $\star$ gates can be "collapsed" into a single $\star$ gate of higher arity, provided that new unary gates can be applied to the inputs.

In [BHH92] it is explained that a unary gate $f_{A}$ is $\times$-collapsible if and only if $A$ is of the form

$$
\left(\begin{array}{ll}
a & 0 \\
0 & b
\end{array}\right) \text { or } \quad\left(\begin{array}{ll}
0 & a \\
b & 0
\end{array}\right),
$$

and + -collapsible if and only if $A$ is of the form

$$
\left(\begin{array}{ll}
a & b \\
0 & c
\end{array}\right)
$$

The following are equivalent definitions of $\star$-collapsible that will be used in this paper.

Property 2 The following are equivalent

1. $f_{A}$ is + -collapsible.
2. $f_{A}(x)=\alpha x+\beta$ for some $\alpha, \beta \in \mathcal{K}$ and $\alpha \neq 0$.
3. $f_{A}(\infty)=\infty$.

The following are equivalent

1. $f_{A}$ is $\times$-collapsible.
2. $f_{A}(x)=\alpha x^{\beta}$ for some $\alpha \in \mathcal{K}$ and $\beta \in\{1, \Leftrightarrow 1\}$.
3. $\left\{f_{A}(\infty), f_{A}(0)\right\}=\{0, \infty\}$.

Proof: We prove the property by showing that $1 \Leftrightarrow 2 \Leftrightarrow 3$. If $f_{A}$ is + -collapsible then

$$
A=\left(\begin{array}{ll}
a & b \\
0 & c
\end{array}\right)
$$

and therefore $f(x)=(a / c) x+(b / c)$. Since $A$ is nonsingular $a \neq 0$ and $c \neq 0$ and $a / c \neq 0.2 \Rightarrow 1$ is obvious. If $f_{A}(x)=\alpha x+\beta$ for some $\alpha, \beta \in \mathcal{K}$ and $\alpha \neq 0$ then $f_{A}(\infty)=\alpha \infty+\beta=\infty$. If $f_{A}(\infty)=\infty$ then since $f_{\left(\begin{array}{ll}a & b \\ d\end{array}\right)}(\infty)=a / d=\infty$ we must have $d=0$.

The result for $\times$-collapsible is left for the reader. $\square$
In [BHH92], a three-way justifying assignment is defined as an assignment of constant values to all but three variables in an AROF such that the resulting formula depends on all of the three remaining variables. For the present results, we require assignments that meet additional requirements, which are defined below.

For any two gates, $\alpha$ and $\beta$, with $\alpha<\beta$, define the $\alpha-\beta$ path as the sequence of gate operations along the path in the tree from $\alpha$ to $\beta$. Define a non-collapsing three-way justifying assignment as a three-way justifying assignment with the following additional property. For the unassigned variables $x, y$, and $z$, if $1 \mathrm{ca}(x, y)<\operatorname{lca}(x, z)$ and all non-unary operations in the lca $(x, y)-\operatorname{lca}(x, z)$ path are of the same type $\star$ (for some $\star \in\{+, \times\}$ ) then the function that results from the justifying assignment is of the form

$$
f_{E}\left(f_{C}\left(f_{A}(x) \star f_{B}(y)\right) \star f_{D}(z)\right)
$$

for some unary operations $f_{A}, f_{B}, f_{C}, f_{D}$ and $f_{E}$, where $f_{C}$ is not $\star$-collapsible. Intuitively, this means that, after the justifying assignment, the two gates, lca $(x, y)$ and $1 \mathrm{ca}(x, z)$, cannot be collapsed-and thus the relationship $1 \mathrm{ca}(x, y)<\operatorname{lca}(x, z)$ can still be detected in the resulting function.

Now, define a total non-collapsing three-way justifying assignment as a single assignment of constant values to all variables in an AROF such that, for any three variables, if all but those three are assigned to their respective constants then the resulting assignment is non-collapsing three-way justifying.

## 5 Parallel Learning Algorithm

In this section, we present a parallel algorithm for learning AROFs. The algorithm has three principal components: finding a total non-collapsing three-way justifying assignment; determining the skeleton of the AROF; and, determining the unary gates of the AROF.

The basic idea is to first construct a graph (that will later be referred to as the LCAH graph) that contains information about the relative positions of the lcas of all pairs of variables. This cannot be obtained quickly in parallel from justifying assignments, because of the possibility that some of the important structure of an AROF "collapses" under any given justifying assignment. However, we shall see
that any total non-collapsing justifying assignment is sufficient to determine the entire structure of the AROF at once (modulo some polylog processing).

Once the LCAH graph has been constructed, the skeleton of the AROF can be constructed by discarding some of the structure of the LCAH graph (a "garbage collection" step). This is accomplished using some simple graph algorithms, as well as a parallel prefix sum computation (which is $\mathrm{NC}^{1}$ computable [LF80]).

Finally, once that skeleton is determined, the unary gates can be determined by a recursive tree contraction method (using results from [B74]).

### 5.1 Finding a Total Non-Collapsing Three-Way Justifying Assignment

In [BHH92], it is proven that, for any triple of variables $x, y$ and $z$, by drawing random values (independently) from a sufficiently large field, and assigning them to the other variables in an AROF, a three-way justifying assignment for those variables is obtained with high probability. In the parallel algorithm, a three-way justifying assignment that is total non-collapsing is required. We show that, if the size of the field $\mathcal{K}$ is at least $O\left(n^{2}\right)$ then the same randomized procedure also yields a total non-collapsing three-way justifying assignment with probability at least $\frac{1}{2}$. Therefore in time $O(1)$ this step can be implemented.

We shall begin with some preliminary lemmas and then the precise statement that we require will appear in Corollary 4.
Lemma 1: If $g(y, z)=f_{B}\left(f_{A}(y) \star z\right)$, where $f_{A}$ is not $\star$-collapsing then there exists at most one value $z^{(0)}$ for $z$ such that $f_{C}(y) \equiv g\left(y, z^{(0)}\right)$ is $\star$-collapsing.
Proof: Let $\star=+$. If $f_{B}\left(f_{A}(y)+z_{0}\right)$ is + -collapsible then by property 2 we have

$$
f_{B}\left(f_{A}(y)+z_{0}\right)=\alpha y+\beta
$$

where $\alpha \in \mathcal{K} \backslash\{0\}$ and $\beta \in \mathcal{K}$. We substitute $y=\infty$ and get

$$
f_{B}\left(f_{A}(\infty)+z_{0}\right)=\infty
$$

Since $f_{A}$ is not + -collapsible, by property 2 , we have $f_{A}(\infty)=\gamma \neq \infty$. Solving the above system using property 1 we get

$$
z_{0}=f_{B^{-1}}(\infty) \Leftrightarrow \gamma
$$

This shows that there is at most one value of $z$ that makes $f_{B}\left(f_{A}(y)+z\right)+-$ collapsible.

Let $\star=\times$. If $f_{B}\left(f_{A}(y) z_{0}\right)$ is $\times$-collapsible then by property 2 we have

$$
f_{B}\left(f_{A}(y)+z_{0}\right)=\alpha y^{\beta},
$$

where $\alpha \in \mathcal{K} \backslash\{0\}$ and $\beta \in\{+1, \Leftrightarrow 1\}$. We substitute $y=0, \infty$ and get

$$
\left\{f_{B}\left(f_{A}(0) z_{0}\right), f_{B}\left(f_{A}(\infty) z_{0}\right)\right\}=\{0, \infty\}
$$

Since $f_{A}$ is not $\times$-collapsible, by property 2 , we have either $f_{A}(0)$ or $f_{A}(\infty)$ is not in $\{0, \infty\}$. Suppose $f_{A}(0) \notin\{0, \infty\}$ and suppose $f_{B}\left(f_{A}(0) z_{0}\right)=0$ (the other cases are similar). Solving this gives

$$
z_{0}=f_{B^{-1}}(0) / f_{A}(0)
$$

This shows that there is at most one value of $z$ that makes $f_{B}\left(f_{A}(y) z\right) \times$-collapsible.
Lemma 2: Let $F\left(x_{1}, \ldots, x_{n}\right)$ be an AROF with lca $\left(x_{1}, x_{2}\right)<\operatorname{lca}\left(x_{1}, x_{3}\right)$ and suppose that all non-unary operations in the lca $\left(x_{1}, x_{2}\right) \Leftrightarrow l c a\left(x_{1}, x_{3}\right)$ path are of the same type $\star \in\{+, \times\}$. Let $x_{4}^{(0)}, \ldots, x_{n}^{(0)}$ be independently uniformly randomly chosen from $S \subseteq \mathcal{K}$, where $|S|=m$. Then the probability that $x_{4}^{(0)}, \ldots, x_{n}^{(0)}$ is a non-collapsing three-way justifying assignment is at least $1 \Leftrightarrow\left(\frac{3 n+1}{m}\right)$.
Proof: Note that $x_{4}^{(0)}, \ldots, x_{n}^{(0)}$ is not a non-collapsing three-way justifying assignment if and only if it is not a justifying assignment or there exists a path between the lcas of $x_{1}, x_{2}$ and $x_{3}$ all non-unary operations are of the same type and the path collapses under the assignment. From [BHH92], the probability of the former condition is at most $\frac{2 n+4}{m}$. We need to bound the probability of the latter condition.

We have that $F\left(x_{1}, \ldots, x_{n}\right)$ is of the form

$$
\left.E\left(f_{H_{k}}\left(\cdots f_{H_{1}}\left(f_{H_{0}}\left(A\left(x_{1}\right) \star B\left(x_{2}\right)\right) \star C_{1}\right) \cdots\right) \star C_{k}\right) \star D\left(x_{3}\right)\right),
$$

where $A\left(x_{1}\right), B\left(x_{2}\right), C_{1}, \ldots, C_{k}, D\left(x_{3}\right), E(y)$ may depend on variables from $x_{4}, \ldots, x_{n}$ in addition to their marked arguments. Let $\bar{A}\left(x_{1}\right), \bar{B}\left(x_{1}\right), \bar{C}_{1}, \ldots, \bar{C}_{k}, \bar{D}\left(x_{3}\right), \bar{E}(y)$ denote the above formulas (respectively) with $x_{4}^{(0)}, \ldots, x_{n}^{(0)}$ substituted for the variables $x_{4}, \ldots, x_{n}$. Also, let $d_{1}, \ldots, d_{k}$ denote the degrees of $C_{1}, \ldots, C_{k}$ (respectively), as functions of $x_{4}, \ldots, x_{n}$. By the assumption that $F$ is in normal form, $f_{H_{0}}$ is not *-collapsing. Therefore, by Lemma 1, there exists at most one value of $C_{1}$ for which $f_{H_{1}}\left(f_{H_{0}}(y) \star C_{1}\right)$ is $\star$-collapsing. Since the degree of $C_{1}$ is $d_{1}$, the probability of this value occurring $C_{1}$ is at most $d_{1} / m$ by Schwartz's result in [Sch80].

Similarly, if $f_{H_{1}}\left(f_{H_{0}}(y) \star C_{1}\right)$ is not $\star$-collapsing then Lemma 1 implies that there exists at most one value of $C_{2}$ for which $f_{H_{2}}\left(f_{H_{1}}\left(f_{H_{0}}(y) \star \bar{C}_{1}\right) \star C_{2}\right)$ is $\star$ collapsing, which occurs with probability at most $d_{2} / m$, and so on. It follows that the probability that

$$
f_{H_{k}}\left(f_{H_{k-1}}\left(\cdots f_{H_{1}}\left(f_{H_{0}}(y) \star \bar{C}_{1}\right) \cdots\right) \star \bar{C}_{k}\right)
$$

is $*$-collapsing is at most $\left(d_{1}+\cdots+d_{k}\right) \frac{1}{m} \leq \frac{n-3}{m}$.
The result now follows by summing the two bounds.
Theorem 3: Let $F\left(x_{1}, \ldots, x_{n}\right)$ be an AROF over $\mathcal{K}$, and $x_{1}^{(0)}, \ldots, x_{n}^{(0)}$ be chosen uniformly from a set $S \subseteq \mathcal{K}$ with $|S|=m$. Then the probability that $x_{1}^{(0)}, \ldots, x_{n}^{(0)}$ is a total non-collapsing three-way justifying assignment is at least $1 \Leftrightarrow \frac{6 n^{2}}{m}$.
Proof: First, note that, from Lemma 2, we can immediately infer that if $x_{1}^{(0)}, \ldots, x_{n}^{(0)}$ are drawn independently uniformly randomly from $S \subseteq \mathcal{K}$, where
$|S|=m$ then the probability that $x_{1}^{(0)}, \ldots, x_{n}^{(0)}$ is a non-collapsing three-way justifying assignment is at most $\binom{n}{3}\left(\frac{n+1}{2 m}\right)=O\left(\frac{n^{4}}{m}\right)$.

To obtain a better bound, consider each subformula $C_{i}$ that is an input to some non-unary gate in the AROF. By results in [BHH92], there are at most two possible values of $C_{i}$ that will result in some triple of variables with respect to which the the assignment is not three-way justifying (the values are 0 and $\infty$ ). Thus, as in the proof of Lemma 2, the probability of one of these values arising for $C_{i}$ is at most $\frac{2 d}{m}$, where $d$ is the degree of $C_{i}$. Also, from Lemma 2, there is at most one value of $C_{i}$ that will result in a collapsing assignment, and the probability of this arising is at most $\frac{d}{m}$. Thus, the probability of one of the two events above arising is at most $\frac{3 d}{m}$, and, since $d \leq n$, this is at most $\frac{3 n}{m}$.

Since there are at most $2 n$ such subformulas $C_{i}$, the probability of any one of them attaining one of the above values is at most $\frac{6 n^{2}}{m}$.

The constant in the proof of theorem 3 can be improved to obtain probability of

$$
1 \Leftrightarrow \frac{\frac{3}{2}\left(n^{2}+3 n \Leftrightarrow 2\right)}{m} .
$$

by using the following observation. Notice that we upper bounded the degree of each subtree by $n$. In fact we can upper bound the degree of the leaves (there are $n$ leaves) by degree 1 since they are variables. Then we have another $n \Leftrightarrow 1$ internal subformulas of degrees $d_{1}<d_{2}<\cdots<d_{n-1}$. It is easy to show that $d_{i} \leq i+1$ (simple induction on the number of nodes). Taking all this into account we obtain the above bound.
By setting $m \geq 3\left(n^{2}+3 n \Leftrightarrow 2\right)$, we obtain the following.
Corollary 4: Let $F\left(x_{1}, \ldots, x_{n}\right)$ be an AROF over $\mathcal{K}$, and $x_{1}^{(0)}, \ldots, x_{n}^{(0)}$ be chosen uniformly from a set $S \subseteq \mathcal{K}$ with $|S|=3\left(n^{2}+3 n \Leftrightarrow 2\right)$. Then the probability that $x_{1}^{(0)}, \ldots, x_{n}^{(0)}$ is a total non-collapsing three-way justifying assignment is least $\frac{1}{2}$.

This Corollary implies that the time complexity of finding an total non-collapsing three-way justifying assignment is $O(1)$.

### 5.2 Determining the Skeleton of a Read-Once Formula in Parallel

In this section, we assume that a total non-collapsing three-way justifying assignment is given and show how to construct the skeleton with $O\left(n^{3}\right)$ membership queries in one parallel step followed by an $O(\log n)$ steps of computation.

Firstly, suppose that, for a triple of variables $x, y$, and $z$, we wish to test whether or not $\operatorname{lca}(x, y)<\operatorname{lca}(x, z)$. If $\operatorname{op}(x, y) \neq \mathrm{op}(x, z)$ then this can be accomplished by a direct application of the techniques in [BHH92], using the fact that we have an assignment that is justifying with respect to variables $x, y$, and $z$. On the other hand, if $\mathrm{op}(x, y)=\mathrm{op}(x, z)$ then $1 \mathrm{ca}(x, y)<1 \mathrm{ca}(x, z)$ could be difficult to detect with a mere justifying assignment because the justifying assignment might
collapse the relative structure between these three variables. If all the non-unary operations in the $\operatorname{lca}(x, y)-1 \mathrm{ca}(x, z)$ path are identical then, due to the fact that we have a non-collapsing justifying assignment, we are guaranteed that the substructure between the three variables does not collapse, and we can determine that $1 \mathrm{ca}(x, y)<1 \mathrm{ca}(x, z)$ in $O(1)$ time (again by directly applying techniques in [BHH92]). The leaves the case where $\operatorname{op}(x, y)=o p(x, z)$ but the non-unary operations in the lca $(x, y)-1 \mathrm{ca}(x, z)$ path are not all of the same type. In this case, the techniques of [BHH92] might fail to determine that $\mathrm{lca}(x, y)<\operatorname{lca}(x, z)$ and report them as equal. We shall overcome this problem at a later stage in our learning algorithm, by making inferences based on hierarcical relationships with other variables. For the time being, we can, in time $O(1)$ with one processor, compute the following.

## DESCENDANT $(x, y, z)$ <br> $$
= \begin{cases}\text { YES } & \text { if } \operatorname{lca}(x, y)<\operatorname{lca}(x, z) \text { and } \operatorname{op}(x, y) \neq \operatorname{op}(x, z) ; \\ \text { YES } & \text { if } \operatorname{lca}(x, y)<\operatorname{lca}(x, z) \text { and all non-unary operations } \\ \text { in the } \operatorname{lca}(x, y)-1 \operatorname{ca}(x, z) \text { path are of the same type; } \\ \text { YES or MAYBE } & \text { if } 1 \operatorname{ca}(x, y)<\operatorname{lca}(x, z) \text { and } \operatorname{op}(x, y)=o p(x, z) \text { but not } \\ \text { all non-unary operations in the lca }(x, y)-1 \operatorname{ca}(x, z) \\ \text { path are of the same type; }\end{cases}
$$

Note that if DESCENDANT $(x, y, z)=$ YES then it must be that $1 \mathrm{ca}(x, y)<$ lca $(x, z)$; however, if DESCENDANT $(x, y, z)=$ MAYBE then it is possible that $\operatorname{lca}(x, y)<\operatorname{lca}(x, z)$, but $\operatorname{op}(x, y)=\operatorname{op}(x, z)$ and the non-unary operations on the $1 \mathrm{ca}(x, y)-\mathrm{lca}(x, z)$ are not of the same type, or that $1 \mathrm{ca}(x, y) \nless 1 \mathrm{ca}(x, z)$.

To construct the extended skeleton of an AROF, we first construct its least common ancestor hierarchy (LCAH) graph, which is defined as follows.

Definition: The least common ancestor hierarchy (LCAH) graph of an AROF with $n$ variables consists of $\binom{n}{2}$ vertices, one corresponding to each (unordered) pair of variables. For the distinct variables, $x$ and $y$, denote the corresponding vertex by $x y$ or, equivalently, $y x$. Then, for distinct vertices $x y$ and $z w$, the directed edge $x y \rightarrow z w$ is present in the LCAH graph if and only if $\operatorname{lca}(x, y) \leq \operatorname{lca}(z, w)$.

We shall prove that the following algorithm constructs the LCAH graph of an AROF.

## Algorithm CONSTRUCT-LCAH-GRAPH

> 1. in parallel for all distinct variables $x, y, z$ do if DESCENDANT $(x, y, z)=$ YES then
> insert edges $x y \rightarrow x z$ and $x y \rightarrow y z$ and $x z \rightarrow y z$ and $y z \rightarrow x z$
2. in parallel for all distinct variables $x, y, z, w$ do
if edges $x y \rightarrow x w \rightarrow x z$ are present then
insert edge $x y \rightarrow x z$

## 3. in parallel for all distinct variables $x, y, z$ do

if no edges between any of $x y, x z, y z$ are present then insert edges in each direction between every pair of $x y, x z, y z$
4. in parallel for all distinct variables $x, y, z, w$ do if edges $x y \rightarrow x w \rightarrow z w$ present or edges $x y \rightarrow y w \rightarrow z w$ present then insert edge $x y \rightarrow z w$

Theorem 5: Algorithm CONSTRUCT-LCAH-GRAPH constructs the LCAH graph of an AROF.

Proof: The proof follows from the following sequence of observations:
(i) For all distinct variables $x, y$ and $z$ for which $\operatorname{lca}(x, y)<\operatorname{lca}(x, z)=\operatorname{lca}(y, z)$, after executing steps 1 and 2 of the algorithm, the appropriate edges pertaining to vertices $x y, x z$ and $y z$ (namely, $x y \rightarrow x z, x y \rightarrow y z, x z \rightarrow y z$ and $y z \rightarrow x z$ ) are present.
(ii) For all distinct variables $x, y$ and $z$ for which $1 \mathrm{ca}(x, y)=\operatorname{lca}(x, z)=1 \mathrm{ca}(y, z)$, after executing step 3 of the algorithm, the appropriate edges pertaining to vertices $x y, x z$ and $y z$ (namely, edges in both directions between every pair) are present.
(iii) For all distinct variables $x, y, z$ and $w$, after executing step 4 of the algorithm, the edge $x y \rightarrow z w$ is present if and only if $\operatorname{lca}(x, y) \leq \operatorname{lca}(z, w)$. $\square$
It is straightforward to verify that algorithm CONSTRUCT-LCAH-GRAPH can be implemented to run in $O(\log n)$ time on an EREW PRAM with $O\left(n^{4}\right)$ processors. Moreover, the $O\left(n^{3}\right)$ membership queries can be made initially in one parallel step.

In an AROF, each non-unary gate corresponds to a biconnected component (which is a clique) of its LCAH graph. Thus, to transform the LCAH graph into the extended skeleton of the AROF, we simply "compress" each of its connected components into a single vertex and then extract the underlying tree structure of this graph (where the underlying tree structure of a graph is the tree whose transitive closure is the graph ${ }^{1}$ ).

This is accomplished using standard graph algorithm techniques, including a parallel prefix sum computation ([LF80]). The details follow.

We first designate a "leader" vertex for each component. We then record the individual variables that are descendants of each non-unary gate, and then discard the other nodes in each connected component.

The algorithm below selects a leader from each connected component in an LCAH graph. We assume that there is a total ordering $\prec$ on the vertices of the LCAH graph (for example, the lexicographic ordering on the pair of indices of the two variables corresponding to each vertex).

## Algorithm LEADER

in parallel for all vertices $x y \prec z w$ do
if edges $x y \rightarrow z w$ and $z w \rightarrow x y$ are present then

[^1]```
mark \(x y\) with X
```

It is easy to prove the following.
Lemma 6: After executing algorithm LEADER, there is precisely one unmarked node (namely, the largest in the $\prec$ ordering) in each connected component of the LCAH graph.

After selecting a leader from each component of the LCAH graph, we add $n$ new nodes to this graph that correspond to the $n$ variables. The edge $x \rightarrow y z$ is inserted if and only if the variable $x$ is a descendant of $\operatorname{lca}(y, z)$. This is accomplished by the following algorithm.

## Algorithm LEAVES

in parallel for all distinct variables $x, y, z, w$ do

$$
\begin{aligned}
& \text { insert edge } x \rightarrow x y \\
& \text { if edge } x y \rightarrow z w \text { is present then } \\
& \text { insert edge } x \rightarrow z w
\end{aligned}
$$

Lemma 7: After executing algorithm LEAVES, the edge $x \rightarrow y z$ in present if and only if variable $x$ is a descendant of lca $(y, z)$.
Both algorithms LEADER and LEAVES can be implemented in $O(1)$ time with $O\left(n^{4}\right)$ processors.

After these steps, the marked nodes are discarded from the augmented LCAH graph (that contains $\binom{n}{2}+n$ vertices), resulting in a graph with at most $2 n \Leftrightarrow 1$ vertices that is isomorphic to the extended skeleton of the AROF. This discarding is accomplished by a standard technique involving the computation of prefix sums. We first adopt the convention that the order $\prec$ extends to the augmented LCAH graph as $x_{1} \prec \cdots \prec x_{n}$ and $x \prec y z$ for any variables $x, y$ and $z$. Then, for each node $v$, set

$$
\varphi(v)= \begin{cases}1 & \text { if } v \text { is unmarked } \\ 0 & \text { if } v \text { is marked }\end{cases}
$$

and compute the prefix sums

$$
\sigma(v)=\sum_{u \preceq v} \varphi(u) .
$$

With algorithms for parallel prefix sum computation ([LF80]) this can be accomplished in $O\left(\log \left(\binom{n}{2}+n\right)\right)=O(\log n)$ time with $O\left(\binom{n}{2}+n\right)=O\left(n^{2}\right)$ processors.

The function $\sigma$ is a bijection between the unmarked nodes of the augmented LCAH graph and some $S \subseteq\{1,2, \ldots, 2 n \Leftrightarrow 1\}$, and $\sigma\left(x_{i}\right)=i$ when $i \in\{1, \ldots, n\}$. The following algorithm uses the values of this function to produce the extended skeleton of the AROF.

## Algorithm COMPRESS-AND-PRUNE

in parallel for all distinct vertices $u, v$ do
if vertices $u, v$ are both unmarked
and edge $u \rightarrow v$ is in augmented LCAH graph then insert edge $\sigma(u) \rightarrow \sigma(v)$ in skeleton graph
in parallel for all distinct $i, j, k \in S$ do
if edges $i \rightarrow j \rightarrow k$ and $i \rightarrow k$ are in skeleton graph then remove edge $i \rightarrow k$ from skeleton graph

The following is straightforward to prove.
Lemma 5: The "skeleton" graph that COMPRESS-AND-PRUNE produces is isomorphic to the extended skeleton of the $A R O F$, where the inputs $x_{1}, \ldots, x_{n}$ correspond to the vertices $1, \ldots, n$ (respectively) of the graph.

### 5.3 Determining a Read-Once Formula from its Skeleton

Once the skeleton of an AROF is determined, what remains is to determine the constants in its unary gates (note that the non-unary operations are easy to determine using the techniques in [BHH92]). We show how to do this in $O\left(\log ^{2} n\right)$ steps with $O(n \log n)$ processors. The main idea is to find a node that partitions the skeleton into three parts whose sizes are all bounded by half of the size of the skeleton. Then the unary gates are determined on each of the parts (in a recursive manner), and the unary gates required to "assemble" the parts are computed.

The following lemma is an immediate consequence from a result in [B74].
Lemma 9 [B74]: For any formula $F\left(x_{1}, \ldots, x_{n}\right)$, there exists a non-unary gate of type $\star$ that "evenly" partitions it in the following sense. With a possible relabelling of the indices of the variables,

$$
F\left(x_{1}, \ldots, x_{n}\right) \equiv G\left(f_{A}\left(f_{B}\left(H\left(x_{1}, \ldots, x_{k}\right)\right) \star f_{C}\left(I\left(x_{k+1}, \ldots, x_{l}\right)\right)\right), x_{l+1}, \ldots, x_{n}\right)
$$

and the number of variables in $G\left(y, x_{l+1}, \ldots, x_{n}\right), H\left(x_{1}, \ldots, x_{k}\right)$ and $I\left(x_{k+1}, \ldots, x_{l}\right)$ are all bounded above by $\left\lceil\frac{n}{2}\right\rceil$.

A minor technicality in the above lemma is that, since the skeleton is not necessarily a binary tree, it may be necessary to "split" a non-binary gate into two smaller gates.

It is straightforward to obtain the above decomposition of a skeleton in $\mathrm{NC}^{1}$. Once this decomposition is obtained, the recursive algorithm for computing the unary gates of the ROF follows from the following lemma.
Lemma 10: Let $x_{1}^{(0)}, \ldots, x_{n}^{(0)}$ be a total non-collapsing justifying assignment for the AROF $F\left(x_{1}, \ldots, x_{n}\right)$. If

$$
\begin{gathered}
F\left(x_{1}, \ldots, x_{n}\right) \equiv G\left(f _ { A } \left(f_{B}\left(H\left(x_{1}, \ldots, x_{k}\right)\right) \star\right.\right. \\
\left.\left.f_{C}\left(I\left(x_{k+1}, \ldots, x_{l}\right)\right)\right), x_{l+1}, \ldots, x_{n}\right)
\end{gathered}
$$

then:
(i) Given the skeleton of $F\left(x_{1}, \ldots, x_{n}\right)$ and the subformulas $G\left(y, x_{l+1}, \ldots, x_{n}\right)$, $H\left(x_{1}, \ldots, x_{k}\right)$ and $I\left(x_{k+1}, \ldots, x_{l}\right)$, it is possible to determine $A, B$ and $C$, and, thus, the entire structure of $F\left(x_{1}, \ldots, x_{n}\right)$ in $O(\log n)$ steps with $O(n \log n)$ processors.
(ii) Given the skeleton of $F\left(x_{1}, \ldots, x_{n}\right)$, the problem of determining $G\left(y, x_{l+1}, \ldots, x_{n}\right)$, $H\left(x_{1}, \ldots, x_{k}\right)$ and $I\left(x_{k+1}, \ldots, x_{l}\right)$ is reducible to the problem of determining a $R O F$ given its skeleton.

Proof: For part (i), assume that the subformulas $G\left(y, x_{l+1}, \ldots, x_{n}\right), H\left(x_{1}, \ldots, x_{k}\right)$ and $I\left(x_{k+1}, \ldots, x_{l}\right)$ are given. Since $x_{1}^{(0)}, \ldots, x_{n}^{(0)}$ is a justifying assignment, $G\left(y, x_{l+1}^{(0)}, \ldots, x_{n}^{(0)}\right)$, $H\left(x_{1}, x_{2}^{(0)}, \ldots, x_{k}^{(0)}\right), I\left(x_{k+1}, x_{k+2}^{(0)}, \ldots, x_{l}^{(0)}\right)$ are all nonconstant unary functions, so there exist nonsingular matrices $A^{\prime}, B^{\prime}, C^{\prime}$ (which are easy to determine in $O(\log n)$ parallel steps) such that

$$
\begin{aligned}
f_{A^{\prime}}(y) & \equiv G\left(y, x_{l+1}^{(0)}, \ldots, x_{n}^{(0)}\right) \\
f_{B^{\prime}}\left(x_{1}\right) & \equiv H\left(x_{1}, x_{2}^{(0)}, \ldots, x_{k}^{(0)}\right) \\
f_{C^{\prime}}\left(x_{k+1}\right) & \equiv I\left(x_{k+1}, x_{k+2}^{(0)}, \ldots, x_{l}^{(0)}\right) .
\end{aligned}
$$

Also,

$$
F\left(x_{1}, x_{2}^{(0)}, \ldots, x_{k}^{(0)}, x_{k+1}, x_{k+2}^{(0)}, \ldots, x_{n}^{(0)}\right) \equiv f_{A^{\prime}}\left(f _ { A } \left(f _ { B } \left(f_{B^{\prime}}\left(x_{1}\right) \star f_{C}\left(f_{C^{\prime}}\left(x_{k+1}\right)\right)\right.\right.\right.
$$

so the matrices $A^{\prime} \cdot A, B \cdot B^{\prime}, C \cdot C^{\prime}$ can be determined in $O(1)$ steps. From this, the matrices $A, B, C$ can be determined.

For part (ii), consider the problem of determining $G\left(y, x_{l+1}, \ldots, x_{n}\right)$. Note that

$$
F\left(y, x_{2}^{(0)}, \ldots, x_{l}^{(0)}, x_{l+1}, \ldots, x_{n}\right) \equiv G\left(f_{A^{\prime \prime}}(y), x_{l+1}, \ldots, x_{n}\right)
$$

for some nonsingular $A^{\prime \prime}$. Therefore, if we fix $x_{2}, \ldots, x_{l}$ to $x_{2}^{(0)}, \ldots, x_{l}^{(0)}$ then we have a reduction from the problem of determining $G\left(f_{A^{\prime \prime}}(y), x_{l+1}, \ldots, x_{n}\right)$.

Similarly, we have reductions from the problem of determining $f_{B^{\prime \prime}}\left(H\left(x_{1}, \ldots, x_{k}\right)\right)$ and $f_{C^{\prime \prime}}\left(I\left(x_{k+1}, \ldots, x_{l}\right)\right)$ for nonsingular matrices $B^{\prime \prime}$ and $C^{\prime \prime}$. Since the matrices $A^{\prime \prime}, B^{\prime \prime}, C^{\prime \prime}$ can be absorbed into the processing of part (i) this is sufficient.

By recursively applying Lemmas 9 and 10 , we obtain a parallel algorithm to determine an AROF given its skeleton and a total noncollapsing three-way justifying assignment in $O\left(\log ^{2} n\right)$ steps. The processor count for this can be bounded by $O(n \log n)$.

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[^1]:    ${ }^{1}$ All edges are directed towards the root.

