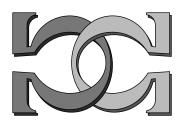




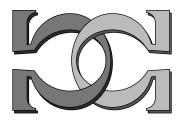
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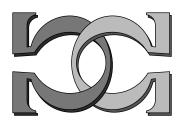
Computing downwards accumulations on trees quickly



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ABSTRACT. Downwards passes on binary trees are essentially functions which pass information down a tree, from the root towards the leaves. Under certain conditions, a downwards pass is both 'efficient' (computable in a functional style in parallel time proportional to the depth of the tree) and 'manipulable' (enjoying a number of distributivity properties useful in program construction); we call a downwards pass satisfying these conditions a downwards accumulation. In this paper, we show that these conditions do in fact yield a stronger conclusion: the accumulation can be computed in parallel time proportional to the logarithm of the depth of the tree, on a CREW PRAM machine.

1 Introduction

The value of programming calculi for the development of correct programs is now clear to the computer science community; their value is even greater for parallel programming than it is for sequential programming, on account of the greater complexity of parallel computations. One such programming calculus is the *Bird-Meertens formalism* (Meertens, 1986; Bird, 1987, 1988; Backhouse, 1989), which relies on the algebraic properties of data structures to provide a body of program transformation rules. This emphasis on the properties of data leads to a 'data parallel' programming style (Hillis and Steele, 1986), which appears to be a promising vehicle for architecture-independent parallel computation (Skillicorn, 1990, 1994).

This paper is concerned with one particular data-parallel operation on one particular data structure, namely *downwards passes* on binary trees. Downwards passes are essentially functions which 'pass information down a tree', from the root towards the leaves. A downwards pass replaces every element of a tree with some function of that element's *ancestors*.

In general, downwards passes are neither 'efficient' (computable in a functional style in parallel time proportional to the product of the depth of the tree and the time taken by the individual operations) nor homomorphic (enjoying certain desirable program transformation properties). However, under certain conditions on the individual operations, downwards passes are both efficient and homomorphic; such downwards passes are called *downwards accumulations*.

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Downwards accumulations, together with their natural counterpart, *upwards accumulations* (Gibbons, 1991, 1993), form the basis of many tree algorithms. For example:

- the *parallel prefix* algorithm (Ladner and Fischer, 1980) is simply an upwards accumulation followed by a downwards accumulation;
- attribute grammars (Knuth, 1968) can be completely evaluated in two passes by performing an upwards followed by a downwards accumulation using 'continuations' (Gibbons, 1991);
- the backwards analysis of a functional program to determine strictness information (Hughes, 1990) is just a downwards accumulation on the parse tree of that program.

The purpose of this paper is to show that the conditions under which downwards passes are efficient and homomorphic, and hence are downwards accumulations, are in fact sufficient to allow them to be computed on a CREW PRAM (but not on a functional machine) in time proportional to the product of the *logarithm* of the depth of the tree and the time taken by the individual operations—which is significantly faster than the obvious way of computing them. This resolves one of the questions posed by Gibbons (1991).

The remainder of this paper is organized as follows. In Section 2, we present our notation. In Sections 3 and 4, we summarize the definitions of homomorphic and efficient downwards passes. In Section 5, we prove a theorem, the *Third Homomorphism Theorem for Paths*, concerning downwards accumulations. Finally, in Section 6, we show that, under certain conditions, a downwards pass can be computed on a CREW PRAM in parallel time proportional to the product of the *logarithm* of the depth of the tree and the time taken by the individual operations. The Third Homomorphism Theorem tells us that, in fact, all homomorphic and efficient downwards passes satisfy these conditions.

2 Notation

We write function composition with an infix 'o':

$$(f \circ g)(a) = f(g(a))$$

We make much use of infix binary operators. Such operators can be turned into unary functions by *sectioning* or partial application:

$$\langle a \oplus \rangle(b) = a \oplus b = \langle \oplus b \rangle(a)$$

Data types are constructed as the 'least solutions' of recursive type equations. The type tree(A) of homogeneous, regular, non-empty binary trees with labels of type A is defined by

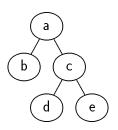
$$tree(A) = Lf(A) + Br(tree(A), A, tree(A))$$

Informally, this says that:

- if a is of type A, then Lf(a) (a leaf labelled with a) is of type tree(A);
- if x and y are of type tree(A) and a is of type A then Br(x, a, y) (a branch labelled with a, with children x and y) is of type tree(A);
- moreover, nothing else is of type tree(A).

For example, the expression

corresponds to the tree



which we call five, and use as an example later.

Homomorphisms form an important class of functions over a given data type. They are the functions that 'promote through' the type constructors. The tree function h is a homomorphism if there is a function g such that

$$h(Br(x, a, y)) = g(h(x), a, h(y))$$

for all x, a and y. In fact, one consequence of the definition of a type as the *least* solution of a type equation is that, for given f and g, there is a unique homomorphism h such that, for all x, a and y, the equations

$$\begin{array}{rcl} h(Lf(a)) &=& f(a) \\ h(Br(x,a,y)) &=& g(h(x),a,h(y)) \end{array}$$

hold. In essence, this solution is a 'relabelling': it replaces every occurrence of Lf in a tree with f, and every occurrence of Br with g.

Homomorphisms are well-behaved, in the sense that they obey a number of 'promotion' or distributivity laws useful for proving properties of programs (Malcolm, 1990). They can also be computed in parallel time proportional to the product of the 'depth' of the structure and the time taken by the individual operations.

One example of a tree homomorphism is the function map(f), which applies f to every element of a tree:

$$\begin{array}{lll} \mathsf{map}(f)(\mathsf{Lf}(a)) &=& \mathsf{Lf}(f(a))\\ \mathsf{map}(f)(\mathsf{Br}(x,a,y)) &=& \mathsf{Br}(\mathsf{map}(f)(x),f(a),\mathsf{map}(f)(y)) \end{array}$$

3 Paths

The definitions and concepts in this section and the next are based, with minor changes, on those of Gibbons (1991). Another presentation is given by Gibbons (1993).

Define the type path(A) as the least solution of the equation

path(A) = Sp(A) + path(A) + path(A) + path(A) + path(A)

modulo some laws described below. That is, for every a of type A, there is a singleton path Sp(a) labelled with a, and for paths x and y there are paths $x \leftrightarrow y$ and $x \leftrightarrow y$. The constructors \leftrightarrow and \leftrightarrow are pronounced 'left turn' and 'right turn' respectively.

The laws obeyed by the path constructors are that \notin and # cooperate with each other—the four equations

$$x \notin (y \notin z) = (x \notin y) \notin z$$

$$x \notin (y \# z) = (x \notin y) \# z$$

$$x \# (y \# z) = (x \# y) \# z$$

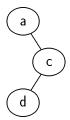
$$x \# (y \# z) = (x \# y) \# z$$

hold. This 'cooperativity property' is a generalization of associativity. It means that any path expression can be written as a sequence of singleton paths joined with # and #, and that parentheses are not needed for disambiguation. Paths are a generalization of non-empty lists, which are defined as the least solution of the equation

$$list(A) = Sl(A) + list(A) + list(A)$$

modulo the law that # is associative. Paths could be thought of as non-empty lists, but with two 'colours' (say, *lemon* and *red*) of concatenation constructor.

We use paths to represent the *ancestors* of an element in a tree. For example, the ancestors of the element d in the tree five form the path



which is represented by the expression Sp(a) # Sp(c) # Sp(d). This correspondence explains the pronunciations 'left turn' and 'right turn'. By the 'top' of a path, we mean the first element (a in this case), and by the 'bottom', we mean the last (d).

Path homomorphisms promote through \notin and \notin :

DEFINITION (1) Function h on paths is (\circledast, \boxdot) -homomorphic iff for all x and y,

$$\begin{array}{rcl} h(x \twoheadleftarrow y) &=& h(x) \circledast h(y) \\ h(x \nrightarrow y) &=& h(x) \circledast h(y) \end{array}$$

Function h is homomorphic iff there exist operators \circledast and \boxdot such that h is (\circledast, \boxdot) -homomorphic.

For example,

$$\mathsf{hom}(\mathsf{f},\circledast,\boxtimes)(\mathsf{Sp}(\mathsf{a}) \twoheadrightarrow \mathsf{Sp}(\mathsf{c}) \twoheadleftarrow \mathsf{Sp}(\mathsf{d})) = \mathsf{f}(\mathsf{a}) \boxtimes \mathsf{f}(\mathsf{c}) \circledast \mathsf{f}(\mathsf{d})$$

One simple example of a path homomorphism is the function length returning the length of a path:

$$length = hom(one, +, +)$$

where, for all **a**,

$$one(a) = 1$$

For example,

$$length(Sp(a) + Sp(c) + Sp(d)) = one(a) + one(b) + one(c) = 3$$

More interesting examples can be constructed.

We note in passing that the components of a homomorphism necessarily respect the cooperativity laws on paths:

THEOREM (3) If h is (\circledast , \boxdot)-homomorphic, then \circledast and \boxdot necessarily cooperate on the range of h—the four equations

hold.

PROOF The proof of the third equation is as follows:

$$\begin{array}{ll} h(x) & \boxplus \ (h(y) \circledast h(z)) \\ & = & \left\{ \begin{array}{l} h \ \mathrm{is} \ (\circledast, \circledast) \mathrm{-homomorphic} \end{array} \right\} \\ & h(x) \boxplus h(y \nleftrightarrow z) \\ & = & \left\{ \begin{array}{l} h \ \mathrm{is} \ (\circledast, \boxplus) \mathrm{-homomorphic} \ \mathrm{again} \end{array} \right\} \\ & h(x \nleftrightarrow (y \twoheadleftarrow z)) \end{array}$$

 $= \begin{cases} \# \text{ and } \# \text{ cooperate } \\ h((x \# y) \# z) \\ = \begin{cases} h \text{ is } (\circledast, \mathbb{E}) \text{-homomorphic, twice } \\ (h(x) \mathbb{E} h(y)) \circledast h(z) \end{cases}$

The other three proofs are similar.

4 Downwards passes

Downwards passes are defined in terms of the ancestors of the elements in a tree. The function **paths** replaces every element of a tree with that element's ancestors: DEFINITION (4) The function **paths** is defined by

For example, paths(five) represents the tree

Downwards passes are functions which 'pass information down a tree'. In other words, each element is replaced with some function of its ancestors. The 'shape' of the tree is unchanged; downwards passes are a shapely operation (Jay and Cockett, 1994).

DEFINITION (5) Functions of the form $map(h) \circ paths$ are called *downwards passes*.

Downwards passes are not necessarily easy to compute, since it is not necessarily possible to 'reuse' the value given to a parent in computing the value given to its children. To address this problem, we isolate a particular kind of path function:

DEFINITION (6) Downwards reduction on a path $dr(f, \oplus, \boxplus)$ satisfies

 $\begin{array}{lll} dr(f,\oplus,\boxplus)(Sp(a)) &=& f(a) \\ dr(f,\oplus,\boxplus)(x \nleftrightarrow y) &=& dr(\langle dr(f,\oplus,\boxplus)(x)\oplus\rangle,\oplus,\boxplus)(y) \\ dr(f,\oplus,\boxplus)(x \nleftrightarrow y) &=& dr(\langle dr(f,\oplus,\boxplus)(x)\boxplus\rangle,\oplus,\boxplus)(y) \end{array}$

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In particular,

$$\begin{aligned} & dr(f, \oplus, \boxplus)(Sp(a) \nleftrightarrow y) = dr(\langle f(a) \oplus \rangle, \oplus, \boxplus)(y) \\ & dr(f, \oplus, \boxplus)(Sp(a) \nleftrightarrow y) = dr(\langle f(a) \boxplus \rangle, \oplus, \boxplus)(y) \end{aligned}$$

and, for example,

$$dr(f,\oplus,\boxplus)(Sp(a) \twoheadrightarrow Sp(c) \twoheadleftarrow Sp(d)) \;\;=\;\; (f(a)\boxplus c) \oplus d$$

Thus,

$$\mathsf{length} = \mathsf{dr}(\mathsf{one}, \odot, \odot)$$

where

 $x \odot a = x + 1$

for all x and a; in general,

$$\mathsf{hom}(\mathsf{f}, \circledast, \circledast) = \mathsf{dr}(\mathsf{f}, \oplus, \boxplus)$$

where

 $\begin{array}{rcl} x \oplus a & = & x \circledast f(a) \\ x \boxplus a & = & x \circledast f(a) \end{array}$

and so all path homomorphisms are downwards reductions (but the converse does not hold).

The downwards passes $map(h) \circ paths$ in which h is a downwards reduction are called efficient downwards passes:

DEFINITION (7) Functions of the form $map(dr(f, \oplus, \boxplus)) \circ paths$ are called *efficient* downwards passes.

Efficient downwards passes are, as the name suggests, cheap to compute, since

$$\begin{array}{ll} \mathsf{map}(\mathsf{dr}(\mathsf{f},\oplus,\boxplus))(\mathsf{paths}(\mathsf{Br}(\mathsf{x},\mathsf{a},\mathsf{y}))) &=& \mathsf{Br}(\mathsf{map}(\mathsf{dr}(\langle\mathsf{f}(\mathsf{a})\oplus\rangle,\oplus,\boxplus))(\mathsf{paths}(\mathsf{x})), \\ && \mathsf{f}(\mathsf{a}), \\ && \mathsf{map}(\mathsf{dr}(\langle\mathsf{f}(\mathsf{a})\boxplus\rangle,\oplus,\boxplus))(\mathsf{paths}(\mathsf{y}))) \end{array}$$

and so can be computed in parallel functional time proportional to the product of the depth of the tree and the time taken by the individual operations.

For example, the function depths, which replaces every element of a tree with its depth in the tree, is defined by

depths = map(length) \circ paths = map(dr(one, $\odot, \odot)) \circ$ paths

where \odot is as defined above. The function **depths** can be computed in parallel functional time proportional to the depth of the tree.

Unfortunately, efficient downwards passes are not in general homomorphic, because the result of applying $map(dr(f, \oplus, \boxplus)) \circ paths$ to the tree Br(x, a, y) depends

 \diamond

on the results of applying different operations, $map(dr(\langle f(a) \oplus \rangle, \oplus, \boxplus)) \circ paths$ and $map(dr(\langle f(a) \boxplus \rangle, \oplus, \boxplus)) \circ paths$, to its children x and y. Therefore, efficient downwards passes do not enjoy the promotion properties alluded to earlier. To remedy this problem, we introduce another class of path function:

DEFINITION (8) Upwards reduction on paths $ur(f, \otimes, \boxtimes)$ satisfies

$$\begin{array}{lll} ur(f,\otimes,\boxtimes)(\mathsf{Sp}(a)) &=& f(a) \\ ur(f,\otimes,\boxtimes)(x \nleftrightarrow y) &=& ur(\langle \otimes ur(f,\otimes,\boxtimes)(y) \rangle,\otimes,\boxtimes)(x) \\ ur(f,\otimes,\boxtimes)(x \nrightarrow y) &=& ur(\langle \boxtimes ur(f,\otimes,\boxtimes)(y) \rangle,\otimes,\boxtimes)(x) \end{array}$$

In particular,

$$\begin{array}{rcl} ur(f,\otimes,\boxtimes)(\mathsf{Sp}(\mathsf{a}) \nleftrightarrow \mathsf{y}) &=& \mathsf{a}\otimes ur(f,\otimes,\boxtimes)(\mathsf{y})\\ ur(f,\otimes,\boxtimes)(\mathsf{Sp}(\mathsf{a}) \nrightarrow \mathsf{y}) &=& \mathsf{a}\boxtimes ur(f,\otimes,\boxtimes)(\mathsf{y}) \end{array}$$

For example,

$$ur(f, \otimes, \boxtimes)(Sp(a) \twoheadrightarrow Sp(c) \twoheadleftarrow Sp(d)) = a \boxtimes (c \otimes f(d))$$

The function length on paths is also an upwards reduction and, in general, all path homomorphisms are upwards reductions (but once more, the converse does not hold).

DEFINITION (9) Functions of the form $map(ur(f, \otimes, \boxtimes)) \circ paths$ are called *homo-morphic downwards passes.* \diamond

Since depth is a path homomorphism, the function depths is a homomorphic downwards pass as well as an efficient downwards pass.

Homomorphic downwards passes satisfy

$$\begin{array}{l} \mathsf{map}(\mathsf{ur}(\mathsf{f},\otimes,\boxtimes))(\mathsf{paths}(\mathsf{Br}(\mathsf{x},\mathsf{a},\mathsf{y}))) \\ = & \mathsf{Br}(\mathsf{map}(\langle\mathsf{a}\otimes\rangle)(\mathsf{map}(\mathsf{ur}(\mathsf{f},\otimes,\boxtimes))(\mathsf{paths}(\mathsf{x}))), \\ & \mathsf{f}(\mathsf{a}), \\ & \mathsf{map}(\langle\mathsf{a}\boxtimes\rangle)(\mathsf{map}(\mathsf{ur}(\mathsf{f},\otimes,\boxtimes))(\mathsf{paths}(\mathsf{y})))) \end{array}$$

and so, as the name suggests, are homomorphic. That is, the result of applying a homomorphic downwards pass to a tree Br(x, a, y) can be computed from the results of applying the same operation to x and to y. Unfortunately, these operations can not in general be computed efficiently—the maps $map(\langle a \otimes \rangle)$ and $map(\langle a \otimes \rangle)$ are expensive to compute. Under what conditions do homomorphic downwards passes coincide with efficient downwards passes?

THEOREM (10) If

$$\mathsf{h} = \mathsf{dr}(\mathsf{f}, \oplus, \boxplus) = \mathsf{ur}(\mathsf{f}, \otimes, \boxtimes)$$

then $map(h) \circ paths$ is both efficient and homomorphic.

THEOREM (11) If

 $\begin{array}{rcl} f(a)\oplus b &=& a\otimes f(b)\\ f(a)\boxplus b &=& a\boxtimes f(b) \end{array}$

and \otimes and \boxtimes cooperate with \oplus and \boxplus , that is,

$$\begin{array}{rcl} \mathbf{a}\otimes(\mathbf{b}\oplus\mathbf{c}) &=& (\mathbf{a}\otimes\mathbf{b})\oplus\mathbf{c}\\ \mathbf{a}\otimes(\mathbf{b}\boxplus\mathbf{c}) &=& (\mathbf{a}\otimes\mathbf{b})\boxplus\mathbf{c}\\ \mathbf{a}\otimes(\mathbf{b}\oplus\mathbf{c}) &=& (\mathbf{a}\otimes\mathbf{b})\oplus\mathbf{c}\\ \mathbf{a}\otimes(\mathbf{b}\boxplus\mathbf{c}) &=& (\mathbf{a}\otimes\mathbf{b})\oplus\mathbf{c} \end{array}$$

then

$$dr(f, \oplus, \boxplus) = ur(f, \otimes, \boxtimes)$$

PROOF The proof is by straightforward induction.

COROLLARY (12) Under the premises of Theorem 11 concerning $f, \oplus, \boxplus, \otimes$ and \boxtimes , the efficient downwards pass $map(dr(f, \oplus, \boxplus)) \circ paths$ is equal to the homomorphic downwards pass $map(ur(f, \otimes, \boxtimes)) \circ paths$.

Thus, under the premises of Theorem 11, we have a downwards pass that is both efficient and homomorphic.

5 The Third Homomorphism Theorem for paths

Recall the data type of non-empty lists mentioned in Section 3. Homomorphisms over such lists are functions h which satisfy

$$h(x + y) = h(x) \circledast h(y)$$

for some associative operator \circledast . Leftwards reductions are functions h which satisfy

$$h(SI(a) + y) = a \oplus h(y)$$

for some (not necessarily associative) \oplus , and rightwards reductions are functions **h** which satisfy

$$h(x + HSI(a)) = h(x) \otimes a$$

for some (again, not necessarily associative) \otimes . Bird's *Third Homomorphism Theorem* on lists (Gibbons, 1996) states that any function which is both a leftwards and a rightwards reduction is also a homomorphism. Thus, for example, any language that is recognizable by both right-to-left and left-to-right sequential algorithms is also recognizable by a 'homomorphic' algorithm, which is much better suited to parallel implementation (Barnard et al., 1991). We show here that a similar theorem holds for paths.

LEMMA (13) For every computable total function h with enumerable domain, there is a computable (but possibly partial) function g such that $h \circ g \circ h = h$.

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PROOF Here is one way of computing g(t) for given t: simply enumerate the domain of h and return the first x such that h(x) = t. If t is in the range of h, \heartsuit then this process terminates.

The path function \mathbf{h} is a homomorphism iff the two implications LEMMA (14)

$$\begin{aligned} h(\mathbf{v}) &= h(\mathbf{x}) \wedge h(\mathbf{w}) = h(\mathbf{y}) \implies h(\mathbf{v} + \mathbf{w}) = h(\mathbf{x} + \mathbf{y}) & --(\mathbf{i}) \\ h(\mathbf{v}) &= h(\mathbf{x}) \wedge h(\mathbf{w}) = h(\mathbf{y}) \implies h(\mathbf{v} + \mathbf{w}) = h(\mathbf{x} + \mathbf{y}) & --(\mathbf{ii}) \end{aligned}$$

hold for all lists v, w, x, y.

PROOF The 'only if' part of the lemma is obvious: if **h** is a homomorphism, then there are operators \circledast and \boxplus such that $h(x + y) = h(x) \circledast h(y)$ and h(x + y) = $h(x) \equiv h(y)$ for all x and y, and the implications trivially hold. Now consider the 'if' part.

Assume that h satisfies (i) and (ii); we must show that h is a homomorphism. Choose a g such that $\mathbf{h} \circ \mathbf{g} \circ \mathbf{h} = \mathbf{h}$, and define operators \circledast and \mathbf{E} by the equations

$$s \circledast t = h(g(s) \notin g(t)) s \circledast t = h(g(s) \# g(t))$$

Because of the way that we chose \mathbf{g} , $\mathbf{h}(\mathbf{x}) = \mathbf{h}(\mathbf{g}(\mathbf{h}(\mathbf{x})))$ and $\mathbf{h}(\mathbf{y}) = \mathbf{h}(\mathbf{g}(\mathbf{h}(\mathbf{y})))$, and so, by (i) (with $\mathbf{v} = \mathbf{g}(\mathbf{h}(\mathbf{x}))$ and $\mathbf{w} = \mathbf{g}(\mathbf{h}(\mathbf{y}))$), we have

$$h(x \nleftrightarrow y) = h(g(h(x)) \nleftrightarrow g(h(y))) = h(x) \circledast h(y)$$

Similarly, by (ii) we have

$$h(x \leftrightarrow y) = h(x) \otimes h(y)$$

and hence **h** is (\circledast, \bigstar) -homomorphic.

THEOREM (15) (Third Homomorphism Theorem for Paths) If

$$\mathsf{h} = \mathsf{dr}(\mathsf{f}, \oplus, \boxplus) = \mathsf{ur}(\mathsf{f}, \otimes, \boxtimes)$$

then h is a path homomorphism.

>

PROOF Suppose $h = dr(f, \oplus, \boxplus) = ur(f, \otimes, \boxtimes)$, h(v) = h(x) and h(w) = h(y). Then

$$\begin{array}{l} h(v \nleftrightarrow w) \\ = & \left\{ \mbox{ since } h = dr(f, \oplus, \boxplus) \end{tabular} \right\} \\ dr(f, \oplus, \boxplus)(v \nleftrightarrow w) \\ = & \left\{ \mbox{ downwards reductions } \right\} \\ dr(\langle dr(f, \oplus, \boxplus)(v) \oplus \rangle, \oplus, \boxplus)(w) \\ = & \left\{ \mbox{ since } h = dr(f, \oplus, \boxplus) \end{tabular} \right\} \\ dr(\langle h(v) \oplus \rangle, \oplus, \boxplus)(w) \end{array}$$

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$$\begin{array}{ll} = & \left\{ \begin{array}{l} {\rm given} \ h(v) = h(x) \end{array} \right\} \\ & dr(\langle h(x) \oplus \rangle, \oplus, \boxplus)(w) \\ = & \left\{ \begin{array}{l} {\rm reversing \ first \ three \ steps \end{array} \right\} \\ & h(x \nleftrightarrow w) \\ = & \left\{ \begin{array}{l} {\rm similarly, \ using \ } h = ur(f, \otimes, \boxtimes) \end{array} \right\} \\ & h(x \nleftrightarrow y) \end{array}$$

Similarly, we get

$$h(v \nleftrightarrow w) = h(x \nleftrightarrow y)$$

Hence, by Lemma 14, h is a homomorphism.

Thus, the conditions under which the downwards pass $map(h) \circ paths$ is efficient and homomorphic—namely, that h is both a downwards and an upwards reduction on paths—are sufficient to ensure that the downwards pass is in fact a path *homomorphism* mapped over the paths of a tree. (Note, however, that the operators involved in the path homomorphism do not necessarily take the same time to compute as those involved in the downwards and upwards reductions.) We therefore choose this as the definition of a downwards accumulation.

DEFINITION (16) Downwards accumulation on trees $da(f, \circledast, \boxdot)$ satisfies

$$da(f, \circledast, \circledast) = map(hom(f, \circledast, \circledast)) \circ paths$$

We show next how to compute such an accumulation in time *logarithmic* in the depth of the tree on a CREW PRAM.

6 Computing downwards accumulations in logarithmic time

Suppose the binary tree has a processor at every node. The processor at node v maintains a pointer v.p, initially to the parent of v. The pointer at the root of the tree is initially nil. The processor at node v also maintains a value v.val; on completion of the algorithm, v.val will hold the result for node v.

We show first how to compute the accumulation $da(f, \circledast, \circledast)$ which, for simplicity, does not differentiate between left and right children. We then modify the algorithm to compute the more general accumulation $da(f, \circledast, \boxtimes)$.

For a node with ancestors $Sp(a) \oplus Sp(c) \oplus Sp(d)$, we have to compute the value $f(a) \oplus f(c) \oplus f(d)$. Every processor v initializes v.val to the result of applying f to v.l, the label of node v. Then we proceed by 'pointer doubling' (Wyllie, 1979): every processor v for which v.p is not nil 'adds' to v.val the val held by processor v.p, then sets v.p to the p held by processor v.p. Initially, every processor holds the 'sum' of just one value, but each iteration doubles the number of values summed,

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so $\lceil \log d \rceil$ iterations suffice to compute the accumulation, where d is the depth of the tree.

The program is as follows:

```
for each node v in parallel do begin
v.val := f(v.l);
while v.p \neq nil do
v.val, v.p := v.p.val \circledast v.val, v.p.p
end
```

The invariant for the inner loop is that, at the start of the ith iteration, $\mathbf{v}.\mathbf{val}$ holds the result of applying $\mathsf{hom}(\mathbf{f}, \circledast, \circledast)$ to the bottom 2^{i-1} elements of the path from the root to \mathbf{v} (or to the whole path, if it has less than 2^{i-1} elements), and $\mathbf{v}.\mathbf{p}$ points to the lowest ancestor not included in this 'sum' (or nil, if all ancestors are included).

Clearly, the inner loop makes at most $\lceil \log d \rceil$ iterations, each of which performs one application of \circledast and a number of pointer manipulations. The whole program takes time proportional to the product of $\lceil \log d \rceil$ and the time taken by \circledast .

The inner loop in this program causes a read conflict. On the first iteration, each parent is asked for its value by both of its children at once; on the second, by each of its (up to) four grandchildren at once; and so on. Hence, this algorithm is not suitable for an EREW PRAM.

We have shown how to compute the downwards accumulation $da(f, \circledast, \circledast)$, in which left and right children are treated the same. It is straightforward to compute the more general accumulation $da(f, \circledast, \boxdot)$. The only difference is that each processor v must record whether it is a left or right descendant of v.p, and perform \circledast or \boxdot accordingly. Each processor v maintains a variable v.s, the 'side', which is initially | for left children and r for right children (and not used for the root). The program is as follows:

```
for each node v in parallel do begin
v.val := f(v.l);
while v.p ≠ nil do
    if v.s = l then
        v.val, v.s, v.p := v.p.val ⊕ v.val, v.p.s, v.p.p
    else
        v.val, v.s, v.p := v.p.val ⊞ v.val, v.p.s, v.p.p
end
```

Thus, the accumulation $da(f, \circledast, \boxdot)$ can be computed on a CREW PRAM in time proportional to the product of the *logarithm* of the depth of the tree and the time taken by the individual \circledast and \boxdot operations.

7 Conclusions

Gibbons (1991) showed that, if f, \oplus and \boxplus permit operators \otimes and \boxtimes satisfying $f(a) \oplus b = a \otimes f(b)$ and $f(a) \boxplus b = a \boxtimes f(b)$ such that \otimes and \boxtimes cooperate with \oplus and \boxplus , then the downwards accumulation $da(f, \oplus, \boxplus)$ is both manipulable and efficiently implementable—in time proportional to the product of the depth of the tree and the time taken by the individual operations—in a functional language. We have shown in Section 5 that these conditions are sufficient to ensure that the function applied to every path in the argument is in fact a path homomorphism. This conclusion led to the algorithm in Section 6, which computes the accumulation on a CREW PRAM in time proportional to the product of the *logarithm* of the depth and the time taken by the individual operations, by a process of 'pointer doubling'.

Gibbons et al. (1994) describe an entirely different algorithm for the same problem, based on parallel tree contraction (Miller and Reif, 1985) rather than on pointer doubling. Their algorithm takes time proportional to the logarithm of the *size* of the tree, as opposed to its depth, and so it is slower in general, but it is suitable for the more restrictive EREW PRAM. Their approach can also be used for computing upwards accumulations, whereas the one presented here can not.

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References

- Roland Backhouse (1989). An exploration of the Bird-Meertens formalism. In International Summer School on Constructive Algorithmics, Hollum, Ameland. STOP project. Also available as Technical Report CS 8810, Department of Computer Science, Groningen University, 1988.
- D. T. Barnard, J. P. Schmeiser, and D. B. Skillicorn (1991). Deriving associative operators for language recognition. Bulletin of the European Association for Theoretical Computer Science, 43:131–139.
- Richard S. Bird (1987). An introduction to the theory of lists. In M. Broy, editor, Logic of Programming and Calculi of Discrete Design, pages 3–42. Springer-Verlag. Also available as Technical Monograph PRG-56, from the Programming Research Group, Oxford University.
- Richard S. Bird (1988). Lectures on constructive functional programming. In Manfred Broy, editor, Constructive Methods in Computer Science, pages 151– 218. Springer-Verlag. Also available as Technical Monograph PRG-69, from the Programming Research Group, Oxford University.
- Jeremy Gibbons, Wentong Cai, and David Skillicorn (1994). Efficient parallel algorithms for tree accumulations. Science of Computer Programming, 23:1–18.
- Jeremy Gibbons (1991). Algebras for Tree Algorithms. D. Phil. thesis, Programming

Research Group, Oxford University. Available as Technical Monograph PRG-94.

- Jeremy Gibbons (1993). Upwards and downwards accumulations on trees. In R. S. Bird, C. C. Morgan, and J. C. P. Woodcock, editors, LNCS 669: Mathematics of Program Construction, pages 122–138. Springer-Verlag. A revised version appears in the Proceedings of the Massey Functional Programming Workshop, 1992.
- Jeremy Gibbons (1996). The Third Homomorphism Theorem. Journal of Functional Programming, 6(4). Earlier version appeared in C. B. Jay, editor, Computing: The Australian Theory Seminar, Sydney, December 1994, p. 62–69.
- W. Daniel Hillis and Guy L. Steele (1986). Data parallel algorithms. Communications of the ACM, 29(12):1170–1183.
- John Hughes (1990). Compile-time analysis of functional programs. In David A. Turner, editor, Research Topics in Functional Programming, pages 117–153. Addison-Wesley.
- C. Barry Jay and J. R. B. Cockett (1994). Shapely types and shape polymorphism. In Donald Sannella, editor, LNCS 788: Programming Languages and Systems—ESOP '94, pages 302–316. Springer-Verlag.
- Donald E. Knuth (1968). Semantics of context-free languages. Mathematical Systems Theory, 2(2):127–145.
- Richard E. Ladner and Michael J. Fischer (1980). Parallel prefix computation. Journal of the ACM, 27(4):831–838.
- Grant Malcolm (1990). Algebraic Data Types and Program Transformation. PhD thesis, Rijksuniversiteit Groningen.
- Lambert Meertens (1986). Algorithmics: Towards programming as a mathematical activity. In J. W. de Bakker, M. Hazewinkel, and J. K. Lenstra, editors, Proc. CWI Symposium on Mathematics and Computer Science, pages 289– 334. North-Holland.
- Gary L. Miller and John H. Reif (1985). Parallel tree contraction and its application. In 26th IEEE Symposium on the Foundations of Computer Science, pages 478–489.
- David B. Skillicorn (1990). Architecture independent parallel computation. IEEE Computer, 23(12):38-51.
- David B. Skillicorn (1994). Foundations of Parallel Programming. Cambridge University Press.
- J. C. Wyllie (1979). The Complexity of Parallel Computations. PhD thesis, Cornell University.