# Deriving tidy drawings of trees 

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#### Abstract

The tree-drawing problem is to produce a 'tidy' mapping of elements of a tree to points in the plane. In this paper, we derive an efficient algorithm for producing tidy drawings of trees. The specification, the starting point for the derivations, consists of a collection of intuitively appealing criteria satisfied by tidy drawings. The derivation shows constructively that these criteria completely determine the drawing. Indeed, the criteria completely determine a simple but inefficient algorithm for drawing a tree, which can be transformed into an efficient algorithm using just standard techniques and a small number of inventive steps.

The algorithm consists of an upwards accumulation followed by a downwards accumulation on the tree, and is further evidence of the utility of these two higher-order tree operations.

Keywords: Derivation, trees, upwards and downwards accumulations, drawing.


## 1 Introduction

The tree drawing problem is to produce a mapping from elements of a tree to points in the plane. This mapping should correspond to a drawing that is in some sense 'tidy'. Our definition of tidiness consists of a collection of intuitively appealing criteria 'obviously' satisfied by tidy drawings.

We derive from these criteria an efficient algorithm for producing tidy drawings of binary trees. The derivation process is a constructive proof that the tidiness criteria completely determine the drawing. In other words, there is only one tidy drawing of any given tree. In fact, the derivation of the algorithm is a completely reasonable and almost routine calculation from the criteria: the algorithm itself, like the drawing, is essentially unique.

[^0]The algorithm that we derive (which is due originally to Reingold and Tilford (1981)) consists of an upwards accumulation followed by a downwards accumulation (Gibbons, 1991; Gibbons, 1993b) on the tree. Basically, an upwards accumulation on a tree replaces every element of that tree with some function of that element's descendents, while a downwards accumulation replaces every element with some function of that element's ancestors. These two higher-order operations on trees are fundamental components of many tree algorithms, such as tree traversals, the parallel prefix algorithm (Ladner and Fischer, 1980), evaluation of attributes in an attribute grammar (Deransart et al., 1988), evaluation of structured queries on text (Skillicorn, 1993), and so on. Their isolation is an important step in understanding and modularizing a tree algorithm. Moreover, work is progressing (Gibbons, 1993a; Gibbons et al., 1994) on the development of efficient parallel algorithms for evaluating upwards and downwards accumulations on a variety of parallel architectures. Identifying the accumulations as components of a known algorithm shows how to implement that algorithm efficiently in parallel.

For the purposes of exposition, we make the simplifying assumption that tree elements are unlabelled or, equivalently, that all labels are the same size. It is easy to generalize the algorithm to cover trees in which the labels may have greatly differing widths. A more interesting generalization covers the case in which tree labels may also have different heights. Bloesch (1993) gives two algorithms for this case. It is slightly more difficult to adapt the algorithm to cope with general trees, in which parents may have arbitrarily but finitely many children. Radack (1988) and Walker (1990) present two different approaches. Radack's algorithm is derived by Gibbons (1991) and described by Kennedy (1995).

The rest of this paper is organized as follows. In Section 2, we briefly describe our notation. In Section 3, we summarize the ideas behind upwards and downwards accumulations on trees. In Section 4, we present the tidiness criteria, and outline a simple but inefficient tree-drawing algorithm. The derivation of an efficient algorithm, the main part of the paper, is in Section 5.

The diagrams in this paper were drawn 'manually' using John Hobby's METAPOST, rather than with the algorithms described here.

## 2 Notation

We will use the Bird-Meertens Formalism or 'BMF' (Backhouse, 1989; Bird, 1987; Bird, 1988; Meertens, 1986), a calculus for the construction of programs from their specifications by a process of equational reasoning. This calculus places great emphasis on notions and properties of data, as opposed to program, structure. The programs we produce are in a functional style, and are readily translated into a modern functional language such as Haskell or ML.

The BMF is known colloquially as 'Squiggol', because its protagonists make heavy use of unusual symbols and syntax. This approach is helpful to the cognoscenti, but tends to make their work appear unnecessarily obscure to the uninitiated. For this reason, we will use a more traditional notation here. We will use mostly words rather
than symbols, and mostly prefix functions rather than infix operators, simply to make expressions easier to parse for those unfamiliar with the calculus. We hasten to add two points. First, this translation leaves the BMF 'philosophy' intact. Second, the presentation here, although more accessible, will be marginally less elegant than it might otherwise have been.

### 2.1 Basic combinators

Sectioning a binary operator involves providing it with one of its arguments, and results in a function of the other argument. For example, $(2+)$ and $(+2)$ are two ways of writing the function that adds two to its argument. The constant function const $a$ returns $a$ for every argument; for example, const $12=1$. (Function application is left-associative, so that this parses as '(const 1) 2', and tightest binding.) Function composition is written ' 0 '; for example, const $1 \circ$ const $2=$ const 1 . The identity function is written ' $i d$ '. The converse $\tilde{\oplus}$ of a binary operator $\oplus$ is obtained by swapping its arguments; for example, $x \sim y=y-x$.

The product type $A \times B$ consists of pairs $(a, b)$ of values, with $a:: A$ and $b:: B$. The projection functions $f s t$ and $s n d$ return the first and second elements of a pair. The fork fork $(f, g)$ of two functions $f$ and $g$ takes a single value and returns a pair; thus, fork $(f, g) a=(f a, g a)$.

### 2.2 Promotion

The notion of promotion comes up repeatedly in the BMF. We say that function $f$ is ' $\oplus$ to $\otimes$ promotable' if, for all $a$ and $b$,

$$
f(a \oplus b)=f a \otimes f b
$$

Promotion is a generalization of distributivity: $f$ distributes through $\oplus \mathrm{iff} f$ is $\oplus$ to $\oplus$ promotable. We say that $f$ 'promotes through $\oplus$ ' if there is a $\otimes$ such that $f$ is $\oplus$ to $\otimes$ promotable.

### 2.3 Lists

The type list $A$ consists of lists of elements of type A. A list is either a singleton [a] for some $a$, or the (associative) concatenation $x+y$ of two lists $x$ and $y$. In this paper, all lists are non-empty. We write 'wrapl' for the function taking $a$ to $[a]$, and write longer lists in square brackets too-for example, ' $[a, b, c]$ ' is an abbreviation for $[a]+[b]+[c]$. For every initial datatype such as lists, there is a higherorder function map, which applies a function to every element of a member of that datatype; for example, map $(+1)[1,2,3]=[2,3,4]$. We will use map for other datatypes such as trees later, and will trust to context to reveal which particular map is meant.

### 2.4 Homomorphisms

An important class of functions on lists are those called homomorphisms. These are the functions that promote through list concatenation. That is, $h$ is a list homomorphism iff there is an associative operator $\otimes$ such that, for all $x$ and $y$,

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

The condition of associativity on $\otimes$ is no great restriction. If $h$ is $\#$ to $\otimes$ promotable then $\otimes$ is necessarily associative, at least on the range of $h$ :

$$
\begin{array}{cc} 
& h x \otimes(h y \otimes h z) \\
= & \{h \text { is }+ \text { to } \otimes \text { promotable }\} \\
= & h x \otimes h(y+z) \\
= & \{\text { promotion again }\} \\
= & h(x+(y+z)) \\
& \{+ \text { is associative }\} \\
= & h((x+y)+z) \\
= & \{\text { promotion, twice }\} \\
& (h x \otimes h y) \otimes h z
\end{array}
$$

In fact, if $h$ is + to $\otimes$ promotable, then it is completely determined by its action on singleton lists; for example,

$$
h[a, b, c]=h([a]+[b]+[c])=h[a] \otimes h[b] \otimes h[c]
$$

If $h$ is $\#$ to $\otimes$ promotable and $h \circ$ wrapl $=f$, then we write $h$ as $l h(f, \otimes)$ (' $l h$ ' stands for 'list homomorphism').

Stated another way, we have the Promotion Theorem on Lists, a special case of the Promotion Theorem (Malcolm, 1990):

Theorem 1
If $h$ is $\oplus$ to $\otimes$ promotable, then

$$
h \circ l h(f, \oplus)=l h(h \circ f, \otimes)
$$

Since $l h($ wrapl,$\#)=i d$, this gives us a vehicle for proving the equality of a function $h$ and a homomorphism $l h(f, \otimes)$, in that we need only show that $h$ is $\#$ to $\otimes$ promotable, and that $h \circ$ wrapl $=f$.

For each $f$, map $f$ is a homomorphism, for

$$
\operatorname{map} f(x+y)=\operatorname{map} f x+\operatorname{map} f y
$$

Indeed, $\operatorname{map} f=l h($ wrapl $\circ f, \#)$, because $\operatorname{map} f[a]=[f a]=($ wrapl $\circ f) a$. Another example of a homomorphism is the function len, which returns the length of a list:

$$
\text { len }=\operatorname{lh}(\text { const } 1,+)
$$

The functions head and last, returning the first and last elements of a list, are also
homomorphisms. For example,

$$
\text { head }(x+y)=\text { head } x=\text { fst }(\text { head } x, \text { head } y)
$$

and so head $=l h(i d, f s t)$. Similarly, last $=l h(i d, s n d)$. Other examples that we will encounter are the functions smallest and largest, which return the smallest and largest elements of a list, respectively:

$$
\begin{aligned}
\text { smallest } & =l h(i d, \text { min }) \\
\text { largest } & =l h(i d, \text { max })
\end{aligned}
$$

and the function sum, which returns the sum of the elements of a list:

$$
\operatorname{sum}=\operatorname{lh}(i d,+)
$$

### 2.5 Leftwards and rightwards functions

Two generalizations of the notion of list homomorphism are the leftwards and the rightwards functions. If there exist $f$ and (not necessarily associative) $\oplus$ such that, for all $a$ and $x$,

$$
\begin{aligned}
h[a] & =f a \\
h([a]+y) & =a \oplus h y
\end{aligned}
$$

then we say that $h$ is leftwards, and write it $l w(f, \oplus)$. Similarly, if for all $x$ and $a$,

$$
\begin{aligned}
h[a] & =f a \\
h(x+[a]) & =h x \otimes a
\end{aligned}
$$

then we say that $h$ is rightwards, and write it $r w(f, \otimes)$. Clearly, if $h$ is a homomorphism then it is both leftwards and rightwards. What is not so obvious is that the converse holds: Bird's Third Homomorphism Theorem (Gibbons, 1993a; Gibbons, 1994c) states that if $h$ is both leftwards and rightwards, then it is a homomorphism.

Consider the function inits, which takes a list and returns the list of lists consisting of its initial segments, in order of increasing length. For example,

$$
\text { inits }[a, b, c]=[[a],[a, b],[a, b, c]]
$$

Now, inits is leftwards, because

$$
\text { inits }([a]+x)=[[a]]+\operatorname{map}([a]+)(\text { inits } x)
$$

In fact,

$$
\text { inits }=l w(\text { wrapl } \circ \text { wrapl } l \oplus) \quad \text { where } \quad a \oplus v=[[a]]+\operatorname{map}([a] \#) v
$$

It is also rightwards, because

$$
\begin{aligned}
\text { inits }(x+[a]) & =\text { inits } x+[x+[a]] \\
& =\text { inits } x+[\text { last (inits } x)+[a]]
\end{aligned}
$$

since last (inits $x)=x$. In fact,

$$
\text { inits }=r w(\text { wrapl } l \circ \text { wrapl }, \otimes) \quad \text { where } \quad w \otimes a=w+[\text { last } w+[a]]
$$



Fig. 1. The tree five

Thus, by the Third Homomorphism Theorem, inits is a list homomorphism.

### 2.6 Binary trees

Finally, we come to binary trees. The type btree $A$ consists of binary trees labelled with elements of type $A$. A binary tree is either a leaf lf a labelled with a single element $a$, or a branch $b r(t, a, u)$ consisting of two children $t$ and $u$ and a label $a$. For example, the expression

$$
b r(l f b, a, b r(l f d, c, l f e))
$$

corresponds to the tree in Figure 1, which we will call five and use as an example later.

Homomorphisms on binary trees $b h(f, \oplus)$ ('binary tree homomorphism') promote through $b r$. That is, they satisfy the equations:

$$
\begin{aligned}
b h(f, \oplus)(l f a) & =f a \\
b h(f, \oplus)(b r(t, a, u)) & =b h(f, \oplus) t \oplus_{a} b h(f, \oplus) u
\end{aligned}
$$

Note that, for binary trees, the second component of a homomorphism is a ternary function. We write its middle argument as a subscript, for lack of anywhere better to put it.

When instantiated to trees, Malcolm's Promotion Theorem states:
Theorem 2
If $h$ satisfies

$$
h(b r(t, a, u))=h t \oplus_{a} h u
$$

then $h=b h(h \circ l f, \oplus)$.
The function map on binary trees satisfies

$$
\begin{align*}
\operatorname{map} f(l f a) & =l f(f a)  \tag{1}\\
\operatorname{map} f(b r(t, a, u)) & =b r(\operatorname{map} f t, f a, \operatorname{map} f u)
\end{align*}
$$

and so

$$
\operatorname{map} f=b h(l f \circ f, \oplus) \quad \text { where } \quad v \oplus_{a} w=b r(v, f a, w)
$$

The function root is a binary tree homomorphism:

$$
\begin{aligned}
\operatorname{root}(l f a) & =a \\
\operatorname{root}(b r(t, a, u)) & =a=\quad \operatorname{root} t \oplus_{a} \operatorname{root} u \quad \text { where } \quad v \oplus_{a} w=a
\end{aligned}
$$

and so, with the same $\oplus$,

$$
\text { root }=b h(i d, \oplus)
$$

So are the functions size and depth:

$$
\begin{aligned}
\text { size } & =b h(\text { const } 1, \oplus) & & \text { where } \quad v \oplus_{a} w=v+1+w \\
\text { depth } & =b h(\text { const } 1, \oplus) & & \text { where } \quad v \oplus_{a} w=1+\max (v, w)
\end{aligned}
$$

and the function brev, which reverses a binary tree:

$$
\text { brev }=b h(l f, \oplus) \quad \text { where } \quad v \oplus_{a} w=b r(w, a, v)
$$

### 2.7 Variable-naming conventions

To help the reader, we make a few conventions about the choice of names. For alphabetic names, single-letter identifiers are typically 'local', their definitions persisting only for a few lines, whereas multi-letter identifiers are 'global', having the same definitions throughout the paper. Elements of lists and trees are denoted $a, b, c, \ldots$. Unary functions are denoted $f, g, h$. Lists and paths (introduced in Section 3.2) are denoted $w, x, y, z$. Trees are denoted $t, u$. The letters $v$ and $w$ are used as the 'results' of functions, for example, in the definitions of homomorphisms such as brev above.

We define a few infix binary operators such as $\oplus$ and $\boxtimes$, just as we might use alphabetic names for variables and unary functions. Round binary operators such as $\oplus$ and $\otimes$ are 'local', and square binary operators such as $\boxplus$ and $\boxtimes$ are 'global'.

## 3 Upwards and downwards accumulations on trees

The material in this section is adapted from (Gibbons, 1993b), which is in turn a summary of (Gibbons, 1991).

### 3.1 Upwards accumulations

Upwards and downwards accumulations arise from considering the list function inits. On trees, the obvious analogue of inits is the function subtrees, which takes a tree and returns a tree of trees. The result is the same shape as the original tree, but each element is replaced by its descendents, that is, by the subtree of the original tree rooted at that element. For example:

$$
\begin{gathered}
\text { subtrees five }=\quad b r(\operatorname{lf}(\operatorname{lf} b), \\
b r(l f), b r(l f d, c, \text { lf } e)), \\
b r(l f(l f d), \\
b r(l f d, c, l f e), \\
l f(l f e)))
\end{gathered}
$$



Fig. 2. The subtrees of five
which corresponds to the tree of trees in Figure 2. The function subtrees is a homomorphism, because it satisfies

$$
\begin{align*}
\text { subtrees }(\text { lf } a) & =\text { lf }(\text { lf } a) \\
\text { subtrees }(b r(t, a, u)) & =b r(\text { subtrees } t, b r(t, a, u), \text { subtrees } u) \tag{2}
\end{align*}
$$

Since root $($ subtrees $t)=t$, we have

$$
\text { subtrees }(b r(t, a, u))=\text { subtrees } t \oplus_{a} \text { subtrees } u
$$

where

$$
v \oplus a w=b r(v, b r(\operatorname{root} v, a, \operatorname{root} w), w)
$$

and so, with the same $\oplus$,

$$
\text { subtrees }=b h(l f \circ l f, \oplus)
$$

The function subtrees replaces every element of a tree with its descendents. An upwards accumulation replaces every element with some function of its descendents. In other words, an upwards accumulation is of the form map $h$ o subtrees for some $h$. In fact, we do not allow $h$ to be an arbitrary function of the descendents. Rather, we insist that $h$ is a tree homomorphism, to ensure that the accumulation can be computed in linear time (assuming that the components of $h$ take constant time). Consider map $h$ (subtrees ( $b r(t, a, u))$ ):

```
    map h(subtrees (br (t,a,u)))
    = {(2)}
    map h(br (subtrees t,br (t,a,u), subtrees u))
    = {(1)}
    br (map h (subtrees t),h(br (t,a,u)), map h(subtrees u))
```

If this is to be computed in linear time, computing $h(b r(t, a, u))$ must take only constant time. If $h=b h(f, \oplus)$ where $f$ and $\oplus$ take constant time, then

$$
h(b r(t, a, u))=h t \oplus_{a} h u
$$



Fig. 3. The path in five to the element labelled $d$
and $h t$ and $h u$ are available in constant time as the roots of map $h$ (subtrees $t$ ) and map $h$ (subtrees $u$ ). Stated another way,

```
map (bh (f,\oplus))\circ subtrees
    = bh(lf\circf,\otimes) where v | | w w br (v, root v}\mp@subsup{\oplus}{a}{}\mathrm{ root }u,u
```

and is therefore both a homomorphism and computable in linear time.
We write 'up $(f, \oplus)$ ' for an upwards accumulation. This satisfies

$$
\begin{equation*}
u p(f, \oplus)=\operatorname{map}(b h(f, \oplus)) \circ \text { subtrees } \tag{3}
\end{equation*}
$$

but, as described above, requires no longer to compute than $b h(f, \oplus)$ does. The function subtrees is itself an upwards accumulation, since subtrees $=$ map id $\circ$ subtrees and $i d$ is a homomorphism; so is $i d$, since $i d=$ map root o subtrees and root is a homomorphism. A more interesting example is the function ndescs, which replaces every element with the number of descendents it has. Letting $\oplus$ satisfy $v \oplus_{a} w=v+1+w$, so that size $=b h($ const $1, \oplus)$, we have

$$
\begin{aligned}
\text { ndescs } & =\operatorname{map}(b h(\text { const } 1, \oplus)) \circ \text { subtrees } \\
& =\text { up }(\text { const } 1, \oplus)
\end{aligned}
$$

Note that the expression involving the map takes quadratic time to compute, whereas the accumulation takes linear time.

### 3.2 Downwards accumulations

Upwards accumulations replace every element of a tree with some function of that element's descendents. For downwards accumulations, on the other hand, we consider an element's ancestors. The ancestors of an element form a path. For example, the ancestors of the element labelled $d$ in five form the path in Figure 3, which could be thought of as a list with two different kinds of concatenation, 'left' and 'right', or as a tree in which each parent has exactly one child. We choose the former view. The type path $A$ consists of paths of elements of type $A$. A path is either a single element $\langle a\rangle$ or two paths $x$ and $y$ joined with a 'left turn', $x \# y$, or a 'right turn', $x \# y$. The function taking $a$ to $\langle a\rangle$ is written 'wrapp'. Just as $\#$ is associative,
the operations $\#$ and $\#$ satisfy the four laws

$$
\begin{aligned}
x+(y+z) & =(x+y)+z \\
x+(y+z) & =(x+y)+z \\
x+(y+z) & =(x+y)+z \\
x+(y+z) & =(x+y)+z
\end{aligned}
$$

We say that ' $H$ cooperates with $H$ ', or ' $\#$ and $H$ cooperate with each other'. Thus, the path in Figure 3 is represented by $\langle a\rangle \#\langle c\rangle \#\langle d\rangle$. Because of the cooperativity property, brackets are unnecessary.

Path homomorphisms promote through both $\#$ and $\#$; if, for all $a, x$ and $y$, the function $h$ satisfies

$$
\begin{aligned}
h\langle a\rangle & =f a \\
h(x+y) & =h x \oplus h y \\
h(x+y) & =h x \otimes h y
\end{aligned}
$$

and $\oplus$ cooperates with $\otimes$, then we write $p h(f, \oplus, \otimes)$ for $h$.
Just as for lists, we generalize path homomorphisms to upwards and downwards functions on paths. If, for all $a, x$ and $y$, the function $h$ satisfies

$$
\begin{aligned}
h\langle a\rangle & =f a \\
h(\langle a\rangle \# y) & =a \oplus h y \\
h(\langle a\rangle H y) & =a \otimes h y
\end{aligned}
$$

then we say that $h$ is upwards, and write it $u w(f, \oplus, \otimes)$. The operators $\oplus$ and $\otimes$ need not enjoy any cooperativity properties. Similarly, if, for all $a, x$ and $y$,

$$
\begin{aligned}
h\langle a\rangle & =f a \\
h(x+\langle a\rangle) & =h x \oplus a \\
h(x+\langle a\rangle) & =h x \otimes a
\end{aligned}
$$

then we say that $h$ is downwards, and write it $d w(f, \oplus, \otimes)$. Path homomorphisms are clearly both upwards and downwards; a generalization of Bird's Third Homomorphism Theorem states the converse.

Theorem 3 (Third Homomorphism Theorem for Paths (Gibbons, 1993a))
A path function that is both upwards and downwards is necessarily a path homomorphism.

The dual for downwards accumulations of the function subtrees is the function paths, which replaces each element of a tree with that element's ancestors. For example:

$$
\begin{aligned}
\text { paths five }=b r & (l f(\langle a\rangle+\langle b\rangle), \\
& \langle a\rangle, \\
& b r(l f(\langle a\rangle+\langle c\rangle+\langle d\rangle), \\
& \langle a\rangle+\langle c\rangle, \\
& l f(\langle a\rangle+\langle c\rangle+\langle e\rangle)))
\end{aligned}
$$

which corresponds to the tree of paths in Figure 4. The function paths is another


Fig. 4. The paths of five
tree homomorphism; it satisfies

$$
\begin{aligned}
\text { paths }(\text { lf } a)= & \text { lf }\langle a\rangle \\
\text { paths }(b r(t, a, u))= & b r(\text { map }(\langle a\rangle+)(\text { paths } t), \\
& \langle a\rangle, \\
& \operatorname{map}(\langle a\rangle+)(\text { paths } u))
\end{aligned}
$$

and so

$$
\text { paths }=b h(l f \circ \text { wrapp }, \oplus)
$$

where

$$
v \oplus_{a} w=b r(\operatorname{map}(\langle a\rangle \#) v,\langle a\rangle, \operatorname{map}(\langle a\rangle \#) w)
$$

A downwards accumulation replaces every element of a tree with some function of that element's ancestors. In other words, downwards accumulations are of the form map $h \circ$ paths for some $h$. Again, we make a restriction on the choice of $h$, but this time it is not so clear just what that restriction should be. On the one hand, we would like $h$ to be upwards, for

$$
\begin{aligned}
& \operatorname{map}(u w(f, \oplus, \otimes))(\text { paths }(b r(t, a, u))) \\
& =b r(\operatorname{map}(a \oplus)(\operatorname{map}(u w(f, \oplus, \otimes))(\text { paths } t)), \\
& \quad f a, \\
& \quad \operatorname{map}(a \otimes)(\operatorname{map}(u w(f, \oplus, \otimes))(\text { paths } u)))
\end{aligned}
$$

and so map $(u w(f, \oplus, \otimes)) \circ$ paths is a homomorphism:

$$
\operatorname{map}(u w(f, \oplus, \otimes)) \circ \text { paths }=b h(l f \circ f, \circledast)
$$

where

$$
v \circledast_{a} w=b r(\operatorname{map}(a \oplus) v, f a, \operatorname{map}(a \otimes) w)
$$

In terms of the Promotion Theorem, this could be stated as follows:
Theorem 4
If

$$
\begin{aligned}
g(l f a) & =l f(f a) \\
g(b r(t, a, u)) & =b r(\operatorname{map}(a \oplus)(g t), f a, \operatorname{map}(a \otimes)(g u))
\end{aligned}
$$

then

$$
g=\operatorname{map}(u w(f, \oplus, \otimes)) \circ \text { paths }
$$

(We will use this theorem later.)
On the other hand, mapping an upwards function over the paths of a tree takes quadratic time to compute, and so we would like $h$ to be downwards, for

$$
\begin{aligned}
& \operatorname{map}(d w(f, \oplus, \otimes))(\text { paths }(b r(t, a, u))) \\
& =\operatorname{br}(\operatorname{map}(d w(((f a) \oplus), \oplus, \otimes))(\text { paths } t), \\
& \quad f a, \\
& \quad \operatorname{map}(d w(((f a) \otimes), \oplus, \otimes))(\text { paths } u))
\end{aligned}
$$

which can be computed in linear time, at the cost of no longer being homomorphic (since the result of applying map $(d w(f, \oplus, \otimes)) \circ$ paths to $b r(t, a, u)$ depends on the results of applying different functions, $\operatorname{map}(d w(((f a) \oplus), \oplus, \otimes)) \circ$ paths and map $(d w(((f a) \otimes), \oplus, \otimes)) \circ$ paths to the children $t$ and $u)$. To satisfy both of these requirements, we insist that $h$ be both upwards and downwards. Theorem 3 concludes that $h$ is therefore a path homomorphism. We write 'down $(f, \oplus, \otimes)$ ' for a downwards accumulation; it satisfies

$$
\begin{equation*}
\text { down }(f, \oplus, \otimes)=\operatorname{map}(p h(f, \oplus, \otimes)) \circ \text { paths } \tag{4}
\end{equation*}
$$

but again can be computed in linear time (if $f, \oplus$ and $\otimes$ each take constant time). Note that $\oplus$ and $\otimes$ must cooperate with each other.

For example, consider the function plen, which returns the length of a path. The function depths replaces every element of a tree with that element's depth in the tree, that is, with the length of its path of ancestors:

$$
\text { depths }=\text { map plen } \circ \text { paths }
$$

As it stands, it is not obvious whether depths is a homomorphism, nor whether it can be computed efficiently. However, plen is upwards,

$$
\text { plen }=u w(\text { const } 1, \oplus, \oplus) \quad \text { where } \quad a \oplus v=1+v
$$

and so depths is a tree homomorphism. Moreover, plen is downwards,

$$
\text { plen }=d w(\text { const } 1, \otimes, \otimes) \quad \text { where } \quad v \otimes a=v+1
$$

and so depths can also be computed in linear time. Writing

$$
\text { depths }=\text { down }(\text { const } 1,+,+)
$$

(since + is associative, it cooperates with itself) shows that depths is both homomorphic and efficiently computable.

We might ask, when can we generalize an upwards function $h$ so that it is also downwards? This would give us an efficient way of computing map $h \circ$ paths.

Suppose $h$ is upwards but not downwards-we cannot write $h(x+\langle a\rangle)$ and $h(x+\langle a\rangle)$ in terms of $h x$ and $a$. Suppose, however, that there is another function $g$ such that $h(x+\langle a\rangle)$ and $h(x+\langle a\rangle)$ can be computed from $h x, g x$ and $a$ : for
some $\ominus$ and $\oslash$,

$$
\begin{aligned}
& h(x+\langle a\rangle)=(h x, g x) \ominus a \\
& h(x+\langle a\rangle)=(h x, g x) \oslash a
\end{aligned}
$$

In a sense, $g$ is the 'extra information' needed to compute $h(x+\langle a\rangle)$ and $h(x+\langle a\rangle)$ from $h x$ and $a$. Now $h$ could be computed downwards, if only we could somehow compute $g$. This, of course, begs the question, how do we compute $g$ ? Suppose further that $g$ is 'self-sustaining', in that no further information is required in order to compute $g$ : for some $\odot$ and $\circledast$,

$$
\begin{aligned}
& g(x+\langle a\rangle)=(h x, g x) \odot a \\
& g(x+\langle a\rangle)=(h x, g x) \circledast a
\end{aligned}
$$

Then fork $(h, g)$ is downwards.
Theorem 5
If

$$
\begin{array}{rlrl}
h\langle a\rangle & =f_{1} a & g\langle a\rangle & =f_{2} a \\
h(x+\langle a\rangle) & =(h x, g x) \ominus a & g(x+\langle a\rangle) & =(h x, g x) \odot a \\
h(x+\langle a\rangle) & =(h x, g x) \oslash a & g(x+\langle a\rangle) & =(h x, g x) \circledast a
\end{array}
$$

then

$$
\text { fork }(h, g)=d w(f, \oplus, \otimes) \quad \text { where } \begin{aligned}
f a & =\left(f_{1} a, f_{2} a\right) \\
(v, w) \oplus a & =(v \ominus a, w \odot a) \\
(v, w) \otimes a & =(v \oslash a, w \circledast a)
\end{aligned}
$$

Then we have $h=f s t \circ$ fork $(h, g)$, and so $h$ is 'almost' downwards-it is the composition of the projection $f s t$ with the downwards function fork $(h, g)$. However, it is not obvious whether fork $(h, g)$ is still upwards. Fortunately, if $g$ is itself upwards, then so is fork $(h, g)$, as shown by the following theorem.
Theorem 6

$$
\text { fork } \begin{aligned}
&\left(u w\left(f_{1}, \ominus, \oslash\right), u w\left(f_{2}, \odot, \circledast\right)\right) \\
&=u w(f, \oplus, \otimes) \text { where } \begin{aligned}
f a & =\left(f_{1} a, f_{2} a\right) \\
& =(a \ominus v, a \odot w) \\
a \oplus(v, w) & =(a \ominus v, a) \\
a \otimes(v, w) & =(a \oslash v, a \circledast w)
\end{aligned}
\end{aligned}
$$

In this case, fork $(h, g)$ is both upwards and downwards, and hence a path homomorphism. Then

$$
\text { maph०paths }=\text { map fst० map }(\text { fork }(h, g)) \circ \text { paths }
$$

which is a (cheap) map composed with a downwards accumulation, and is efficiently computable.

## 4 Drawing binary trees tidily

In this section, we define 'tidiness' and specify the function $b d r a w$, which draws a binary tree. We make the simplifying assumption that all tree labels are the same
size, because, for the purposes of positioning the elements of the tree, we can then ignore the labels altogether.

The first property that we observe of tidy drawings is that all of the elements at a given depth in a tree have the same $y$-coordinate in the drawing. That is, the $y$-coordinate is determined completely by the depth of an element, and the problem reduces to that of finding the $x$-coordinates. This gives us the type of bdraw, the function which draws a binary tree-its argument is of type btree $A$ for some $A$, and its result is a binary tree labelled with $x$-coordinates:

$$
\text { bdraw } \quad:: \quad \text { btree } A \rightarrow \text { btree } \mathbb{D}
$$

where coordinates range over $\mathbb{D}$, the type of distances. We require that $\mathbb{D}$ include the number 1, and be closed under subtraction (and hence also under addition) and halving. Sets satisfying these conditions include the reals, the rationals, and the rationals with finite binary expansions, the last being the smallest such set. We exclude discrete sets such as the integers, as Supowit and Reingold (1983) have shown that the problem is NP-hard with such coordinates.

Tidy drawings are also regular, in the sense that the drawing of a subtree is independent of the context in which it appears. Informally, this means that the drawings of children can be committed to (separate pieces of) paper before considering their parent. The drawing of the parent is then constructed by translating the drawings of the children. In symbols:

$$
b d r a w(b r(t, a, u))=b r(\operatorname{map}(+r)(b d r a w t), b, \text { map }(+s)(b d r a w u))
$$

for some $b, r$ and $s$.
Tidy drawings also exhibit no left-to-right bias. In particular, a parent should be centred over its children. We also specify that the root of a tree should be given $x$-coordinate 0 . Hence, $r+s$ and $b$ in the above equation should both be 0 , as should the position given to the only element of a singleton tree:

$$
\begin{aligned}
b d r a w(l f a) & =l f 0 \\
\text { bdraw }(b r(t, a, u)) & =b r(\text { map }(-s)(\text { bdraw } t), 0, \text { map }(+s)(b d r a w u))
\end{aligned}
$$

for some $s$. Indeed, a tidy drawing will have the left child to the left of the right child, and so $s>0$.

This lack-of-bias property implies that a tree and its mirror image produce drawings which are reflections of each other. That is, if we write '-' for unary negation $\dagger$, then we also require

$$
\text { bdraw } \circ \text { brev }=\text { map }-\circ \text { brev } \circ b d r a w
$$

The fourth criterion is that, in a tidy drawing, elements do not collide, or even get too close together. That is, pictures of children do not overlap, and no two elements on the same level are less than one unit apart.
$\dagger$ The presence of sectioning means that, strictly speaking, we should distinguish between the number 'minus one', written ' -1 ', and the function 'minus one', written ' $(-1)$ '.


Fig. 5. Drawings $p i c_{1}$ and $p i c_{2}$, for which $p i c_{1} \boxplus p i c_{2}=-2$

Finally, a tidy drawing should be as narrow as possible, given the above constraints. Supowit and Reingold (1983) show that narrowness and regularity cannot be satisfied together-there are trees whose narrowest drawings can only be produced by drawing identical subtrees with different shapes-and so one of the two criteria must be made subordinate to the other. We choose to retain the regularity property, since it will lead us to a homomorphic solution.

These last two properties determine $s$, the distance through which children are translated. That distance should be the smallest distance that does not cause violation of the fourth criterion. Suppose the operator $\boxplus$, when given two drawings of trees, returns the width of the narrowest part of the gap between the trees. (If the drawings overlap, this distance will be negative.) For example, if $p i c_{1}$ and $p i c_{2}$ are as in Figure 5, then $p i c_{1} \boxplus p i c_{2}=-2$, the minimum of $0-0,-\frac{1}{2}-\frac{1}{2}$ and $-1-1$. The drawings should be moved apart or together to make this distance 1, that is,

$$
s=(1-(\text { bdraw } t \boxplus \text { bdraw } u)) \div 2
$$

(In the example above, $s$ will be $1 / 2$.)
All that remains to be done to complete the specification is to formalize this description of $\boxplus$.

### 4.1 Levelorder traversal

We define two different 'zip' operators, each of which takes a pair of lists and returns a single list by combining corresponding elements in some way. These two operators are 'short zip', which we write szip, and 'long zip', written lzip. These operators differ in that the length of the result of a short zip is the length of its shorter argument, whereas the length of the result of a long zip is the length of its longer argument. For example:

$$
\begin{aligned}
\operatorname{szip}(\oplus)([a, b],[c, d, e]) & =[a \oplus c, b \oplus d] \\
\operatorname{lzip}(\oplus)([a, b],[c, d, e]) & =[a \oplus c, b \oplus d, e]
\end{aligned}
$$

From the result of the long zip, we see that the $\oplus$ must have type $A \times A \rightarrow A$. This is not necessary for short zip, but we do not use the general case.

The two zips are given formally by the equations

$$
\begin{aligned}
\text { szip }(\oplus)([a],[b]) & =[a \oplus b] \\
\text { szip }(\oplus)([a],[b]+y) & =[a \oplus b] \\
\text { szip }(\oplus)([a]+x,[b]) & =[a \oplus b] \\
\text { szip }(\oplus)([a]+x,[b]+y) & =[a \oplus b]+\text { szip }(\oplus)(x, y) \\
\text { lzip }(\oplus)([a],[b]) & =[a \oplus b] \\
\text { lzip }(\oplus)([a],[b]+y) & =[a \oplus b]+y \\
\text { lzip }(\oplus)([a]+x,[b]) & =[a \oplus b]+x \\
\text { lzip }(\oplus)([a]+x,[b]+y) & =[a \oplus b]+l z i p(\oplus)(x, y)
\end{aligned}
$$

They share many properties, but we use two in particular.

Fact 7
Both szip $(\oplus)(x, y)$ and lzip $(\oplus)(x, y)$ can be evaluated using just min (len $x$, len $y$ ) applications of $\oplus$.

## Lemma 8

If $f$ is $\oplus$ to $\otimes$ promotable, then map $f$ is both szip $(\oplus)$ to szip $(\otimes)$ and lzip $(\oplus)$ to lzip $(\otimes)$ promotable.

We use long zip to define levelorder traversal of binary trees. This is given by the function levels :: btree $A \rightarrow$ list (list $A$ ):

$$
\text { levels }=b h(\text { wrapl } \circ \text { wrapl }, \oplus) \quad \text { where } \quad x \oplus_{a} y=[[a]]+\operatorname{lzip}(+)(x, y)
$$

For example, the levelorder traversals of lf $b$ and $b r$ (lf $d, c, l f e$ ) are $[[b]]$ and $[[c],[d, e]]$, respectively, and so

## levels five

$=[[a]]+\operatorname{lzip}(+)([[b]],[[c],[d, e]])$
$=[[a]]+[[b]+[c],[d, e]]$
$=[[a],[b, c],[d, e]]$
We can at last define the operator $\boxplus$ on pictures, in terms of levelorder traversal. It is given by

$$
\begin{aligned}
p \boxplus q= & \text { smallest }(\text { szip }(\sim) \\
( & \text { map largest }(\text { levels } p), \\
& \text { map smallest }(\text { levels } q)))
\end{aligned}
$$

If $v$ and $w$ are levels at the same depth in $p$ and $q$, then largest $v$ and smallest $w$ are the rightmost point of $v$ and the leftmost point of $w$, respectively, and so smallest $w$ - largest $v$ is the width of the gap at this level. Clearly, $p \boxplus q$ is the minimum over all levels of these gap widths. For example, with $p i c_{1}$ and $p i c_{2}$ as in Figure 5, we have

$$
\begin{aligned}
\text { map largest }\left(\text { levels pic } c_{1}\right) & =[0,1 / 2,1] \\
\text { map smallest }\left(\text { levels pic } c_{2}\right) & =[0,-1 / 2,-1]
\end{aligned}
$$

and so

$$
\text { pic }_{1} \boxplus \text { pic }_{2}=\text { smallest }[0-0,-1 / 2-1 / 2,-1-1]=-2
$$

This completes the specification of $\boxplus$, and hence of bdraw:

$$
\begin{equation*}
\text { bdraw }=b h(\text { const }(\text { lf } 0), \boxtimes) \tag{5}
\end{equation*}
$$

where

$$
\begin{gathered}
p \boxtimes_{a q}=\operatorname{br}(\operatorname{map}(-s) p, 0, \text { map }(+s) q) \quad \text { where } \quad s=(1-(p \boxplus q)) \div 2 \\
p \boxplus q=\operatorname{smallest}(\operatorname{szip}(\underset{\sim}{\sim})(\text { map largest }(\text { levels } p), \\
\text { map smallest }(\text { levels } q)))
\end{gathered}
$$

This specification is executable, but requires quadratic effort. We now derive a linear algorithm to satisfy it.

## 5 Drawing binary trees efficiently

A major source of inefficiency in the program that we have just developed is the occurrence of the two maps in the definition of $\square$. Intuitively, we have to shift the drawings of two children when assembling the drawing of their parent, and then shift the whole lot once more when drawing the grandparent. This is because we are computing directly the absolute position of every element. If instead we were to compute the relative position of each parent with respect to its children, these repeated translations would not occur. A second pass-a downwards accumulation-can fix the absolute positions by accumulating relative positions.

Suppose the function rootrel on drawings of trees satisfies

$$
\begin{aligned}
\text { rootrel }(\text { lf } a) & =0 \\
\text { rootrel }(b r(t, a, u)) & =(a-\operatorname{root} t) \odot(\text { root } u-a)
\end{aligned}
$$

for some idempotent operator $\odot$. The idea here is that rootrel determines the position of a parent relative to its children, given the drawing of the parent. For example, with $p i c_{1}$ as in Figure 5, we have:

$$
\text { rootrel pic } c_{1}=(0--1 / 2) \odot(1 / 2-0)=1 / 2
$$

That is, if we define the function sep by

$$
\begin{equation*}
\text { sep }=\text { rootrel } \circ \text { bdraw } \tag{6}
\end{equation*}
$$

then

$$
\begin{align*}
\operatorname{sep}(l f a) & =0 \\
\operatorname{sep}(b r(t, a, u)) & =(1-(b d \text { raw } t \boxplus \text { bdraw } u)) \div 2 \tag{7}
\end{align*}
$$

For example:

$$
\begin{aligned}
\text { sep five } & =(1-(b d r a w(\text { lf } b) \boxplus \text { bdraw }(\text { br }(\text { lf } d, c, \text { lf } e)))) \div 2 \\
& =(1-0) \div 2 \\
& =1 / 2
\end{aligned}
$$

Then

$$
\begin{gathered}
b d \operatorname{raw}(b r(t, a, u))=b r(\operatorname{map}(-s)(b d r a w t), 0, \operatorname{map}(+s)(b d r a w u)) \\
\text { where } s=\operatorname{sep}(b r(t, a, u))
\end{gathered}
$$

Now, applying sep to each subtree gives the relative (to its children) position of every parent. Define the function rel by

$$
\begin{equation*}
\text { rel }=\text { map sep } \circ \text { subtrees } \tag{8}
\end{equation*}
$$

From this, we calculate that

```
    rel (lf a)
= {(8)}
    map sep (subtrees (lf a))
= {(2)}
    map sep (lf (lf a))
= {(1)}
        lf (sep (lf a))
= {(7)}
        lf 0
```

and

$$
\begin{aligned}
& \operatorname{rel}(b r(t, a, u)) \\
= & \{(8)\} \\
= & \operatorname{map} \operatorname{sep}(\text { subtrees }(b r(t, a, u))) \\
= & \{(2)\} \\
= & \operatorname{map} \operatorname{sep}(b r(\text { subtrees } t, b r(t, a, u), \text { subtrees } u)) \\
= & \{(1)\} \\
= & b r(\text { map sep }(\text { subtrees } t), \operatorname{sep}(b r(t, a, u)), \text { map sep }(\text { subtrees } u))) \\
= & \{(8)\} \\
& b r(\operatorname{rel} t, \operatorname{sep}(b r(t, a, u)), \text { rel } u)
\end{aligned}
$$

That is,

$$
\begin{align*}
\operatorname{rel}(l f a) & =l f 0 \\
\operatorname{rel}(b r(t, a, u)) & =b r(\text { rel } t, \operatorname{sep}(b r(t, a, u)), \operatorname{rel} u) \tag{9}
\end{align*}
$$

This gives us the first 'pass', computing the position of every parent relative to its children. How can we get from this to the absolute position of every element? We need a function abs satisfying the condition

$$
\begin{equation*}
a b s \circ r e l=b d r a w \tag{10}
\end{equation*}
$$

We can calculate from this requirement a definition of $a b s$. On leaves, the condition reduces to

$$
a b s(\operatorname{rel}(l f a))=b d r a w(l f a)
$$

$$
\Leftrightarrow \begin{gathered}
\{(9),(5)\} \\
\\
\text { abs }(\text { lf } 0)=l f 0
\end{gathered}
$$

while on branches we require

$$
\begin{array}{cc} 
& a b s(\operatorname{rel}(b r(t, a, u)))=b d r a w(b r(t, a, u)) \\
\Leftrightarrow & \quad\{(9),(5) ; \text { let } s=\operatorname{sep}(b r(t, a, u))\} \\
& a b s(b r(\text { rel } t, s, r e l u))=b r(\operatorname{map}(-s)(b d r a w t), 0, \text { map }(+s)(b d r a w u)) \\
\Leftrightarrow & \quad\{\operatorname{assuming}(10) \text { holds on smaller trees }\} \\
& a b s(b r(\text { rel } t, s, \text { rel } u))=b r(\operatorname{map}(-s)(a b s(\text { rel } t)), 0, \operatorname{map}(+s)(a b s(\text { rel } u)))
\end{array}
$$

These requirements are satisfied if

$$
\begin{aligned}
a b s(l f a) & =l f 0 \\
a b s(b r(t, a, u)) & =b r(\operatorname{map}(-a)(a b s t), 0, \operatorname{map}(+a)(a b s u))
\end{aligned}
$$

By Theorem 4, this implies that

$$
a b s=\operatorname{map}(u w(\text { const } 0, \stackrel{\sim}{-},+)) \circ \text { paths }
$$

We give the upwards function $u w$ (const $0, \sim,+$ ) a name, pabs ('the absolute position of the bottom of a path'), for brevity:

$$
\text { pabs }=u w(\text { const } 0, \tilde{\sim},+)
$$

so that

$$
\begin{equation*}
a b s=\text { map pabs } \circ \text { paths } \tag{11}
\end{equation*}
$$

Thus, we have

$$
\begin{equation*}
b d r a w=a b s \circ r e l \tag{12}
\end{equation*}
$$

where

$$
\begin{aligned}
\text { rel } & =\text { map sep } \circ \text { subtrees } \\
a b s & =\text { map pabs } \circ \text { paths }
\end{aligned}
$$

This is still inefficient, as computing rel takes quadratic time (because sep is not a tree homomorphism) and computing abs takes quadratic time (because pabs is not path homomorphism). We show next how to compute rel and abs quickly.

### 5.1 An upwards accumulation

We want to find an efficient way of computing the function rel satisfying

$$
\text { rel }=\text { map sep } \circ \text { subtrees }
$$

where

$$
\begin{aligned}
\operatorname{sep}(l f a) & =0 \\
\operatorname{sep}(b r(t, a, u)) & =(1-(\text { bdraw } t \boxplus \text { bdraw } u)) \div 2
\end{aligned}
$$

We have already observed that rel is not an upwards accumulation, because sep is not a homomorphism-more information than the separations of the grandchildren
is needed in order to compute the separation of the children. How much more information is needed? It is not hard to see that, in order to compute the separation of the children, we need to know the 'outlines' of their drawings.

Each level of a picture is sorted. Therefore,

$$
\begin{aligned}
\text { map smallest } \circ \text { levels } & =\text { map head } \circ \text { levels } \\
\text { map largest } \circ \text { levels } & =\text { map last } \circ \text { levels }
\end{aligned}
$$

and so

$$
\begin{equation*}
p \boxplus q=\text { right } p \boxtimes \text { left } q \tag{13}
\end{equation*}
$$

where

$$
\begin{aligned}
\text { left } & =\text { map head } \circ \text { levels } \\
\text { right } & =\text { map last } \circ \text { levels }
\end{aligned}
$$

and

$$
v \boxtimes w=\text { smallest }(\operatorname{szip}(\tilde{-})(v, w))
$$

Intuitively, left and right return the 'contours' of a drawing. For example, applying the function fork (left, right) to the tree pic $c_{1}$ in Figure 5 produces the pair of lists ( $\left[0,-\frac{1}{2}, 0\right],\left[0, \frac{1}{2}, 1\right]$ ). These contours are precisely the extra information needed to make sep a homomorphism.

To show this, we need to show first that sep can be computed from the contours, and second that computing the contours is a homomorphism. Define the function contours by

$$
\begin{equation*}
\text { contours }=\text { fork }(\text { left, right }) \circ b d r a w \tag{14}
\end{equation*}
$$

How do we find sep $t$ from contours $t$ ? By definition, the head of each contour is 0 , and (if $t$ is not just a leaf) the second elements in the contours are - (sep $t$ ) and sep $t$. Thus,

$$
\begin{equation*}
\text { sep }=\text { spread } \circ \text { contours } \tag{15}
\end{equation*}
$$

where, for some idempotent $\odot$,

$$
\begin{aligned}
\text { spread }([0],[0]) & =0 \\
\operatorname{spread}([0]+x,[0]+y) & =-(\text { head } x) \odot \text { head } y
\end{aligned}
$$

on pairs of lists, each with head 0 .
Now we show that contours is a homomorphism. On leaves, we have

$$
\begin{aligned}
& \text { contours (lf a) } \\
& =\{(14)\} \\
& \text { fork (left, right) (bdraw (lf a)) } \\
& =\quad\{(5)\} \\
& \text { fork (left, right) (lf 0) } \\
& =\quad\{\text { left, right }\} \\
& \text { ([0], [0]) }
\end{aligned}
$$

For branches, we will consider just the left contour, as the right contour is sym-
metric. We have

Similarly,

$$
\begin{array}{r}
\text { right }(b d r a w(b r(t, a, u))) \\
=[0]+\text { lzip snd }(\operatorname{map}(-s)(\text { right }(\text { bdraw } t)), \\
\operatorname{map}(+s)(\text { right }(\text { bdraw } u)))
\end{array}
$$

Now,

```
    bdraw t \boxplus bdraw u
= {(13)}
    right (bdraw t)\boxtimes left (bdraw u)
= {(14)}
    snd (contours t)\boxtimesfst (contours u)
```

and so

$$
\text { contours }(b r(t, a, u))=\text { contours } t \boxminus_{a} \text { contours } u
$$

where

$$
\begin{align*}
&(w, x) \boxminus_{a}(y, z) \\
&=([0]+\text { lzip fst }(\operatorname{map}(-s) w, \operatorname{map}(+s) y), \\
& {[0]+\text { lzip snd }(\operatorname{map}(-s) x, \operatorname{map}(+s) z)) }  \tag{16}\\
& \quad \text { where } s=(1-(x \boxtimes y)) \div 2
\end{align*}
$$

Hence,

$$
\begin{equation*}
\text { contours }=b h(\text { const }([0],[0]), \boxminus) \tag{17}
\end{equation*}
$$

Thus,

$$
\begin{aligned}
& \text { rel } \\
= & \{(8)\} \\
= & \text { map sep } \circ \text { subtrees } \\
= & \{(15)\} \\
& \text { map spread } \circ \text { map contours } \circ \text { subtrees } \\
= & \{(17)\} \\
= & \operatorname{map} \text { spread } \circ \text { map }(b h(\text { const }([0],[0]), \boxminus)) \circ \text { subtrees } \\
= & \{(3)\} \\
& \operatorname{map} \text { spread } \circ \text { up }(\text { const }([0],[0]), \boxminus)
\end{aligned}
$$

That is,

$$
\begin{equation*}
\text { rel }=\text { map spread } \circ \text { up }(\text { const }([0],[0]), \boxminus) \tag{18}
\end{equation*}
$$

This is now an upwards accumulation, but it is still expensive to compute. The operation $\boxminus$ takes at least linear effort, resulting in quadratic effort for the upwards accumulation. One further step is needed before we have an efficient algorithm for rel.

We have to find an efficient way of evaluating the operator $\boxminus$ from (16):

$$
\begin{aligned}
&(w, x) \boxminus_{a}(y, z)=([0]+\text { lzip fst }(\operatorname{map}(-s) w, \operatorname{map}(+s) y), \\
& {[0] \text { + lzip snd }(\text { map }(-s) x, \operatorname{map}(+s) z)) } \\
& \text { where } s=(1-(x \boxtimes y)) \div 2
\end{aligned}
$$

One way of doing this is with a data refinement whereby, instead of maintaining a list of absolute distances, we maintain a list of relative distances. That is, we make a data refinement using the invertible abstraction function msi $=$ map sum $\circ$ inits, which computes absolute distances from relative ones. Under this refinement, the maps can be performed in constant time, since

```
map (+s)(msi x) = msi (mapplus (s,x))
    where mapplus (b,[a])=[b+a]
                        mapplus (b,[a]+x)=[b+a]+x
```

Moreover, the zips can still be performed in time proportional to their shorter argument, since if len $x \geq$ len $y$ then

$$
\text { lzip fst }(\text { msi } x, m s i y)=m \text { si } x
$$

and if len $x<$ len $y$ then, letting $\left(y_{1}, y_{2}\right)=\operatorname{split}($ len $x, y)$ where

$$
\begin{aligned}
\text { split }(1,[a]+x) & =([a], x) \\
\operatorname{split}(n+1,[a]+x) & =([a]+v, w) \quad \text { where } \quad(v, w)=\operatorname{split}(n, x)
\end{aligned}
$$

we have

$$
\begin{aligned}
& \text { lzip fst (msi } x \text {, msi } y \text { ) } \\
& =\quad\left\{\text { msi } y=\text { msi } y_{1}+\operatorname{map}\left(+ \text { sum } y_{1}\right)\left(\text { msi } y_{2}\right) \text {; len } x=\text { len } y_{1}\right\} \\
& \text { msi } x+\operatorname{map}\left(+ \text { sum } y_{1}\right)\left(\text { msi } y_{2}\right) \\
& =\{\text { map }(+\operatorname{sum} x) \circ \operatorname{map}(- \text { sum } x)=i d\} \\
& \text { msi } x+\operatorname{map}(+ \text { sum } x)\left(\text { map }\left(- \text { sum } x+\text { sum } y_{1}\right)\left(\text { msi } y_{2}\right)\right) \\
& =\{(19)\} \\
& \text { msi } x+\text { map }(+ \text { sum } x)\left(\text { msi (mapplus }\left(\text { sum } y_{1}-\text { sum } x, y_{2}\right)\right) \text { ) } \\
& =\quad\{\text { msi }(x+y)=\text { msi } x+\operatorname{map}(+ \text { sum } x)(\text { msi } y)\} \\
& \text { msi }\left(x+\text { mapplus }\left(\text { sum } y_{1}-\text { sum } x, y_{2}\right)\right)
\end{aligned}
$$

By symmetry,

$$
\text { lzip snd }(\text { msi } x, \text { msi } y)=\text { lzipfst }(\text { msi } y, m s i x)
$$

(Note that the guard len $x \geq$ len $y$ must also be evaluated in time proportional to the lesser of len $x$ and len $y$, and so cannot be done simply by computing the two lengths. In Figure 6 we define the predicate nst (for 'no shorter than'), for which nst $(x, y)=($ len $x \geq$ len $y)$ but which takes time proportional to the lesser of len $x$ and len $y$.)

The refined $\boxminus$ still takes linear effort because of the zips, but the important observation is that it now takes effort proportional to the length of its shorter argument (that is, to the lesser of the common lengths of $w$ and $x$ and the common lengths of $y$ and $z$, when $\boxminus$ is 'called' with arguments $(w, x)$ and $(y, z)$ ). Reingold and Tilford (1981) show that, if evaluating $h t \oplus_{a} h u$ from $a, h t$ and $h u$ takes effort proportional to the lesser of the depths of the trees $t$ and $u$, then the tree homomorphism $h=b h(f, \oplus)$ can be evaluated with linear effort. Actually, what they show is that if $g$ satisfies

$$
\begin{aligned}
g(l f a) & =0 \\
g(b r(t, a, u)) & =g t+\min (\text { depth } t, \text { depth } u)+g u
\end{aligned}
$$

then

$$
g x=\text { size } x-\text { depth } x
$$

which can easily be proved by induction. Intuitively, $g$ counts the number of pairs of horizontally adjacent elements in a tree.

With this data refinement, rel can be computed in linear time.

### 5.2 A downwards accumulation

We now have an efficient algorithm for rel. All that remains to be done is to find an efficient algorithm for $a b s$, where

$$
\begin{aligned}
a b s & =\text { map pabs } \circ \text { paths } \\
\text { pabs } & =\text { uw (const } 0, \sim,+ \text { ) }
\end{aligned}
$$

We note first that computing $a b s$ as it stands is inefficient. No operator $\oplus$ can satisfy $a+$ const $0 b=$ const $0 a \oplus b$ for all $a$ and $b$, and so pabs cannot be computed downwards, and abs is not a downwards accumulation. Intuitively, pabs starts at the bottom of a path and discards the bottom element, but we cannot do this when starting at the top of the path.

What extra information do we need in order to be able to compute pabs downwards? It turns out that

$$
\begin{align*}
& \text { pabs }(x+\langle a\rangle)=\text { pabs } x-\text { bottom } x  \tag{20}\\
& \text { pabs }(x+\langle a\rangle)=\text { pabs } x+\text { bottom } x
\end{align*}
$$

where bottom returns the bottom element of a path:

$$
\text { bottom }=u w(i d, \text { snd }, \text { snd })
$$

Now, pabs and bottom together can be computed downwards, because of (20) and

$$
\begin{aligned}
\operatorname{bottom}(x+\langle a\rangle) & =a \\
\operatorname{bottom}(x+\langle a\rangle) & =a
\end{aligned}
$$

Let

$$
\begin{equation*}
\text { pabsb }=\text { fork }(\text { pabs, bottom }) \tag{21}
\end{equation*}
$$

Then, by Theorem 6, pabsb is upwards:

$$
\begin{aligned}
& \text { pabsb }=u w(f, \oplus, \otimes) \quad \text { where } \quad f a=(0, a) \\
& a \oplus(v, w)=(v-a, w) \\
& a \otimes(v, w)=(v+a, w)
\end{aligned}
$$

Moreover, by Theorem 5, pabsb is downwards:

$$
\text { pabsb =dw(f,Ф,囚) where } \begin{aligned}
f a & =(0, a) \\
(v, w) \oplus a & =(v-w, a) \\
(v, w) \otimes a & =(v+w, a)
\end{aligned}
$$

Finally, by Theorem 3, pabsb is a path homomorphism:

$$
\begin{align*}
\text { pabsb }=p h(f, \oplus, \otimes) & f \\
\text { where } & =(0, a)  \tag{22}\\
(v, w) \oplus(x, y) & =(v-w+x, y) \\
(v, w) \otimes(x, y) & =(v+w+x, y)
\end{align*}
$$

Putting all this together gives us

$$
\begin{aligned}
& a b s \\
&=\quad\{(11)\}
\end{aligned}
$$

```
    map pabs o paths
= \{ ( 2 1 ) \}
    map fst\circ map pabsb\circ paths
= {(22), with f,\oplus and \otimes as defined there}
    map fst\circ map (ph(f,\oplus,\otimes))\circ paths
= {(4)}
    map fst ० down (f,\oplus,\otimes)
```

That is,

$$
\begin{equation*}
a b s=\text { map fst } \circ \text { down }(f, \oplus, \otimes) \tag{23}
\end{equation*}
$$

which can be computed in linear time.

### 5.3 The program

To summarize, the program that we have derived is as in Figure 6.

## 6 Conclusion

### 6.1 Summary

We have presented a number of natural criteria satisfied by tidy drawings of unlabelled binary trees. From these criteria, we have derived an efficient algorithm for producing such drawings.

The steps in the derivation were as follows:

1. we started with an executable specification (5)-an 'obviously correct' but inefficient program;
2. we eliminated one source of inefficiency, by computing first the position of every parent relative to its children, and then fixing the absolute positions in a second pass (12);
3. we made a step towards making the first pass efficient, by turning the function computing relative positions into an upwards accumulation (18), computing not just relative positions but also the outlines of the drawings;
4. we made a data refinement on the outline of a drawing (19), allowing us to shift it in constant time; and
5. we made the second pass efficient by turning the function computing absolute positions into a downwards accumulation (23), computing not just the absolute positions but also the bottom element of every path. (In fact, we could have calculated, using the technique of strengthening invariants (Gries, 1982) and no invention at all, that

$$
\text { fork }(p a b s, u w(i d, \sim,+))
$$

$$
\begin{aligned}
& \text { bdraw }=a b s \circ r e l \\
& \text { rel }=\text { map spread } \circ \text { up }(\text { const }([0],[0]), \boxminus) \\
& (w, x) \boxminus_{a}(y, z)=([0]+\text { lzipfst ( mapplus }(-s, w) \text {, mapplus }(s, y)), \\
& \text { [0] + lzipsnd (mapplus }(-s, x) \text {, mapplus }(s, z)) \text { ) } \\
& \text { where } \quad s=(1-(x \boxtimes y)) \div 2 \\
& \text { mapplus }(b,[a])=[a+b] \\
& \text { mapplus }(b,[a]+x)=[a+b]+x \\
& \text { lzipfst }(x, y)=x, \quad \text { if } n s t(x, y) \\
& =x+\text { mapplus }(\text { sum } v-\text { sum } x, w) \text {, otherwise } \\
& \text { where }(v, w)=\operatorname{split}(\text { len } x, y) \\
& \text { lzipsnd }(x, y)=\text { lzipfst }(y, x) \\
& \text { nst }(x,[b])=\text { true } \\
& \text { nst }([a],[b]+y)=\text { false } \\
& n s t([a]+x,[b]+y)=n s t(x, y) \\
& \operatorname{split}(1,[a]+x)=([a], x) \\
& \operatorname{split}(n+1,[a]+x)=([a]+v, w) \quad \text { where }(v, w)=\operatorname{split}(n, x) \\
& \text { spread }([0],[0])=0 \\
& \text { spread }([0]+x,[0]+y)=-(\text { head } x) \odot \text { head } y \quad \text { where } \quad a \odot a=a \\
& v \boxtimes w=\operatorname{lh}(i d, \min )(\operatorname{szip}(\sim)(v, w)) \\
& a b s=\text { map fst } \circ \text { down }(f, \oplus, \otimes) \\
& \text { where } \quad f a=(0, a) \\
& (v, w) \oplus(x, y) \quad=(v-w+x, y) \\
& (v, w) \otimes(x, y)=(v+w+x, y)
\end{aligned}
$$

Fig. 6. The final program
is downwards, and hence also a path homomorphism; this would have done just as well.)

The derivation showed several things:

1. the criteria uniquely determine the drawing of a tree;
2. the criteria also determine an inefficient algorithm for drawing a tree (step 1 in the derivation), and only three or four small inventive steps (steps 2 to 5 in the derivation) are needed to transform this into an efficient algorithm;
3. the algorithm (due to Reingold and Tilford (1981)) is just an upwards accumulation followed by a downwards accumulation, and is further evidence of the utility of these higher-order operations;
4. identifying these accumulations as major components of the algorithm may lead, using known techniques for computing accumulations in parallel, to an optimal parallel algorithm for drawing unlabelled binary trees.

### 6.2 Related work

The problem of drawing trees has quite a long and interesting history. Knuth (1968; 1971) and Wirth (1976) both present simple algorithms in which the $\boldsymbol{x}$-coordinate of an element is determined purely by its position in inorder traversal. Wetherell and Shannon (1979) first considered 'aesthetic criteria', but their algorithms all produce biased drawings. Independently of Wetherell and Shannon, Vaucher (1980) gives an algorithm which produces drawings that are simultaneously biased, irregular, and wider than necessary, despite his claims to have 'overcome the problems' of Wirth's simple algorithm. Reingold and Tilford (1981) tackle the problems in the algorithms of Wetherell and Shannon and of Vaucher, by proposing the criteria concerning bias and regularity. Their algorithm is the one derived for binary trees here. Supowit and Reingold (1983) show that it is not possible to satisfy regularity and minimal width simultaneously, and that the problem is NP-hard when restricted to discrete (for example, integer) coordinates. Brüggemann-Klein and Wood (1990) implement Reingold and Tilford's algorithm as macros for the text formatting system $\mathrm{T}_{\mathrm{E}} \mathrm{X}$.

The problem of drawing general trees has had rather less coverage in the literature. General trees are harder to draw than binary trees, because it is not so clear what is meant by 'placing siblings as close as possible'. For example, consider a general tree with three children, $t, u$ and $v$, in which $t$ and $v$ are large but $u$ relatively small. It is not sufficient to consider just adjacent pairs of siblings when spacing the siblings out, because $t$ may collide with $v$. Spacing the siblings out so that $t$ and $v$ do not collide allows some freedom in placing $u$, and care must be taken not to introduce any bias. Reingold and Tilford (1981) mention general trees in passing, but make no reference to the difficulty of producing unbiased drawings. Bloesch (1993) (who adapts the algorithms of Vaucher and of Reingold and Tilford to cope with node labels of varying width and height) also does not attempt to produce unbiased drawings, despite his claims to the contrary. Radack (1988) effectively constructs two drawings, one packing siblings together from the left and the other from the right, and then averages the results. That algorithm is derived by Gibbons (1991) and described by Kennedy (1995). Walker (1990) uses a slightly different method; he positions children from left to right, but when a child touches against a left sibling other than the nearest one, the extra displacement is apportioned among the intervening siblings.

### 6.3 Further work

Gibbons (1991) extends this derivation to general trees. We have yet to apply the methods used here to Bloesch's algorithm (Bloesch, 1993) for drawing trees in which the labels may have different heights, but do not expect it to yield any surprises. It may also be possible to apply the techniques in (Gibbons et al., 1994) to yield an optimal parallel algorithm to draw a binary tree of $n$ elements in $\log n$ time on
$n / \log n$ processors, even when the tree is unbalanced-although this is complicated by having to pass non-constant-size contours around in computing $\boxminus$.

We are currently exploring the application to graphs of some of the general notions-homomorphisms and accumulations-used here on lists and trees. See (Gibbons, 1994b) for further details.

### 6.4 Acknowledgements

Thanks are due to Sue Gibbons and the anonymous referees, whose suggestions improved the presentation of this paper considerably.

## 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.
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. Springer-Verlag. Also available as Technical Monograph PRG-69, from the Programming Research Group, Oxford University.
Anthony Bloesch (1993). Aesthetic layout of generalized trees. Software—Practice and Experience, 23(8):817-827.
Anne Brüggemann-Klein and Derick Wood (1990). Drawing trees nicely with $T_{E} X$. In Malcolm Clark, editor, $T_{E} X$ : Applications, Uses, Methods, pages 185-206. Ellis Horwood.
Pierre Deransart, Martin Jourdan, and Bernard Lorho (1988). LNCS 323: Attribute Grammars-Definitions, Systems and Bibliography. Springer-Verlag.
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 (1993a). Computing downwards accumulations on trees quickly. In Gopal Gupta, George Mohay, and Rodney Topor, editors, 16th Australian Computer Science Conference, pages 685-691, Brisbane. Revised version submitted for publication.
Jeremy Gibbons (1993b). 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 (1994a). How to derive tidy drawings of trees. In C. Calude, M. J. J. Lennon, and H. Maurer, editors, Proceedings of Salodays in Auckland, pages 53-73, Department of Computer Science, University of Auckland.
Jeremy Gibbons (1994b). An initial-algebra approach to directed acyclic graphs. Department of Computer Science, University of Auckland. Accepted for publication in Mathematics of Program Construction 1995.

Jeremy Gibbons (1994c). The Third Homomorphism Theorem. In C. Barry Jay, editor, Computing: The Australian Theory Seminar. University of Technology, Sydney. Submitted for publication.
David Gries (1982). A note on a standard strategy for developing loop invariants and loops. Science of Computer Programming, 2:207-214.
Andrew Kennedy (1995). Drawing trees. Journal of Functional Programming, to appear.
Donald E. Knuth (1968). The Art of Computer Programming, Volume 1: Fundamental Algorithms. Addison-Wesley.
Donald E. Knuth (1971). Optimum binary search trees. Acta Informatica, 1:14-25.
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.
G. M. Radack (1988). Tidy drawing of $M$-ary trees. Technical Report CES-88-24, Department of Computer Engineering and Science, Case Western Reserve University, Cleveland, Ohio.
Edward M. Reingold and John S. Tilford (1981). Tidier drawings of trees. IEEE Transactions on Software Engineering, 7(2):223-228.
David B. Skillicorn (1993). Parallel evaluation of structured queries in text. Draft, Department of Computing and Information Sciences, Queen's University, Kingston, Ontario.
Kenneth J. Supowit and Edward M. Reingold (1983). The complexity of drawing trees nicely. Acta Informatica, 18(4):377-392.
Jean G. Vaucher (1980). Pretty-printing of trees. Software—Practice and Experience, 10:553-561.
John Q. Walker, II (1990). A node-positioning algorithm for general trees. SoftwarePractice and Experience, 20(7):685-705.
Charles Wetherell and Alfred Shannon (1979). Tidy drawings of trees. IEEE Transactions on Software Engineering, 5(5):514-520.
Niklaus Wirth (1976). Algorithms + Data Structures = Programs. Prentice Hall.


[^0]:    $\dagger$ Partially supported by University of Auckland Research Committee grant number A18/XXXXX/62090/3414013.

