# Generic functional programming with types and relations 

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#### Abstract

A generic functional program is one which is parameterised by datatype. By installing specific choices, for example lists or trees, different programs are obtained that are, nevertheless, abstractly the same. The purpose of this paper is to explore the possibility of deriving generic programs. Part of the theory of lists that deals with segments is recast as a theory about 'segments' in a wide class of datatypes, and then used to pose and solve a generic version of a well-known problem.


## Capsule Review

There are many programming problems that can be specified on a variety of datatypes such as lists, trees and forests. Examples of such problems are parsing, pretty printing, unification, etc. This paper investigates how datatype independent programs for problems such as these can be constructed.

The main problem solved in the paper is a generalisation of the maximum segment sum problem on the datatype of lists to a large class of datatypes. Bird's paper 'An introduction to the theory of lists' from 1987 contains an elegant derivation of a linear-time program for the maximum segment sum problem on lists. In this paper, the authors show that this derivation can be generalised in a relatively straightforward way to a large class of datatypes. The main ingredients of the generalisation are functors, describing the structure of datatypes, and relations, simplifying the derivation. The constituents of the derivation are general combinators that are not only applicable to the maximum segment sum problem, but also to problems like deforestation and pattern matching.

## 1 Introduction

To what extent is it possible to construct useful programs without knowing exactly what datatypes are involved? At first sight this may seem a strange question, but take the example of pattern matching. Over lists this problem can be formulated in terms of two strings, a pattern and a text; the object is to determine if and where the pattern occurs as a segment of the text. Now, pattern matching can be generalised to other datatypes, including arrays and trees of various kinds; the essential step is to be able to define the notion of segment in these types. So the intriguing question
arises: can one construct a useful algorithm, parameterised by a datatype, to solve the general problem of pattern matching?

The purpose of this paper is to explore the ideas of generic programming through a second problem, related to but simpler than pattern matching, namely the problem of computing the maximum segment sum. This problem was chosen because sufficient list theory already exists (Bird, 1987, 1989a, 1990) for one to calculate an efficient solution in a few equational steps. It turns out that the theory of segments can be generalised to a wide class of datatypes, so the calculation leads to a generic solution to the problem.

To be able to construct a generic theory of segments, we need a reformulation of the theory of lists with two new ingredients. The first ingredient is a categorical treatment of datatypes (Malcolm, 1990; Manes and Arbib, 1986; Lehmann and Smyth, 1981). In the categorical approach, datatypes are characterised in terms of certain mappings, called functors, and specifications can be parameterised by functors in a simple and direct manner. Although the categorical approach is becoming familiar to functional programmers (Barr and Wells, 1990; Bird and de Moor, 1996; Pierce, 1991), we will give a brief but hopefully adequate account of the essential ideas.

The second ingredient involves the move from functions to relations (Aarts et al., 1992; De Moor, 1992). Introducing relations enables us to deal more smoothly with nondeterministic specifications, but it also turns out that the calculus of relations leads to substantial simplifications both in the derivation of purely functional programs and in the study of general datatypes. Again, we will give a light account of the ideas.

The rest of the paper is structured as follows. In the next section we show quite informally - how the notion of segment can be defined in one or two other data types. After that, in section 3, we examine the structure of the derivation of the maximum segment sum problem, recalling that it depends on two results in the theory of lists, namely Horner's rule and the Scan lemma. Section 4 gives an account of the general theory of datatypes, and in section 5 we review part of the calculus of relations. Using this theory, we show in section 6 how Horner's rule can be generalised. To give an account of a general version of the Scan lemma, we need the fact, explained in section 7, that every datatype found in functional programming comes equipped with a membership relation. It turns out that the existence of a membership relation for a datatype is crucial for generic programming. In section 8 we generalise the Scan lemma, and in section 9 present the complete derivation of the maximum segment sum problem. Finally, Section 10 contains a discussion of the implications of this research.

## 2 Towards generality

Let us start by being more precise about what we mean by a segment of a list, indeed, what we mean by a list. There are two basic views of lists, one of which is
given by the type declaration

$$
\text { listl } A::=\text { nil } \mid \operatorname{snoc}(\text { list } A, A)
$$

Formally, this means that lists are represented as finite terms over nil and snoc. For instance, the list [1,2,3] is represented by the term
snoc (snoc (snoc (nil, 1), 2), 3).

Thinking of lists as terms, we see that a prefix of $x$ is really the same thing as a subterm of $x$. The function subterms takes a list and returns the set of all its subterms:

$$
\begin{aligned}
\text { subterms nil } & =\{\text { nil }\} \\
\text { subterms }(\text { snoc }(x, a)) & =\text { subterms } x \cup\{\operatorname{snoc}(x, a)\} .
\end{aligned}
$$

In the theory of lists, prefixes are called initial segments, and the function subterms is called inits. There is the subtle difference that inits returns a list rather than a set, but we ignore this distinction for now, though we return to it in the next section, and it will prove to be of crucial importance later on.

Dual to the notion of prefix is that of a suffix. A suffix of $x$ can be obtained by substituting the empty list for a subterm of $x$. For instance, snoc (snoc (nil, 2), 3) is obtained from the term above by replacing the subterm snoc (nil, 1) by the empty list nil. For the sake of a word we can say that this subterm is the result of pruning the original term. The function prunings takes a list and returns all ways in which it can be pruned:

$$
\begin{aligned}
\text { prunings nil } & =\{\text { nil }\} \\
\text { prunings }(\operatorname{snoc}(x, a)) & =\{\text { nil }\} \cup\{\operatorname{snoc}(y, a) \mid y \in \text { prunings } x\} .
\end{aligned}
$$

In the theory of lists, suffixes are called tail segments and prunings is called tails.
One can now define arbitrary segments by the equation

$$
\text { segments }=\text { union } \cdot \text { map prunings } \cdot \text { subterms } .
$$

Here union is the function that takes a collection of sets and returns its union, and map is the operator that applies a function to all elements of a set.

For comparison, consider now the other - and more familiar - view of lists, given by the type declaration

$$
\operatorname{listr} A \quad::=\text { nil } \mid \operatorname{cons}(A, \operatorname{listr} A)
$$

With this datatype the role of inits and tails are reversed: subterms gives the tail segments of a list, while prunings gives the initial segments. The function segments is defined in the same way as before and again gives the segments of a list.

As a third example, consider binary trees as defined by

$$
\text { tree } A::=\text { nil } \mid \text { fork }(A, \text { tree } A, \text { tree } A)
$$

The elements of this type are finite trees, this time over nil and fork, so it is again possible to define the functions subterms and prunings. The function subterms takes
a binary tree and returns the set of all its subtrees:

$$
\begin{aligned}
\text { subterms nil } & =\{\text { nil }\} \\
\text { subterms }(\text { fork }(a, x, y)) & =\text { subterms } x \cup \text { subterms } y \cup\{\text { fork }(a, x, y)\} .
\end{aligned}
$$

The function prunings takes a binary tree and substitutes nil for its subtrees in all possible ways:

$$
\begin{aligned}
\text { prunings nil }= & \{\text { nil }\} \\
\text { prunings }(\text { fork }(a, x, y))= & \{\text { nil }\} \cup \\
& \{\text { fork }(a, u, v) \mid u \in \text { prunings } x, v \in \text { prunings } y\} .
\end{aligned}
$$

The segments of a tree are defined by the same equation as before. Jeuring (1989) also considered such a definition, though he spoke of treecuts rather than segments.

## 3 The maximum segment sum

The problem of the maximum segment sum is to compute the function mss, where

$$
\text { mss }=\text { max } \cdot \text { map sum } \cdot \text { segments } .
$$

Over lists this problem is interpreted as follows: given a list of integers, compute the sum of the elements in each segment of the list and return the maximum such sum. For example, mss $[-1,2,-1,3,-2]=4$ because the segment $[2,-1,3]$ has maximum sum.

Given the definition of segments in the previous section, we can calculate

$$
\begin{aligned}
& \text { mss } \\
&=\{\text { definition }\} \\
& \text { max } \cdot \text { map sum } \cdot \text { segments } \\
&= \quad\{\text { definition of segments }\} \\
& \text { max } \cdot \text { map sum } \cdot \text { union } \cdot \text { map prunings } \cdot \text { subterms } \\
&= \quad\{\text { since map } f \cdot \text { union }=\text { union } \cdot \text { map }(\text { map } f)\} \\
& \text { max } \cdot \text { union } \cdot \text { map }(\text { map sum }) \cdot \text { map prunings } \cdot \text { subterms } \\
&=\quad\{\text { since max } \cdot \text { union }=\text { max } \cdot \text { map max } \text { (over sets of non-empty sets })\} \\
& \text { max } \cdot \text { map max } \cdot \text { map }(\text { map sum }) \cdot \text { map prunings } \cdot \text { subterms } \\
&=\quad \quad\{\text { since map } f \cdot \text { map } g=\text { map }(f \cdot g)\} \\
& \max \cdot \text { map }(\text { max } \cdot \text { map sum } \cdot \text { prunings }) \cdot \text { subterms. }
\end{aligned}
$$

In the fourth step we have assumed that prunings returns a non-empty set, so that map prunings subterms returns a set of non-empty sets; this is necessary since max is not defined on the empty set. So far, the calculation has been completely general (and also standard: we have merely copied from Bird (1989) modulo some changes in names), but now let us revert to the specific datatype listl $A$. In this datatype the function sum is defined by

$$
\operatorname{sum}=\text { foldl }(0,+),
$$

where fold $(c, f)$ is defined by

$$
\begin{aligned}
\text { foldl }(c, f) \text { nil } & =c \\
\text { foldl }(c, f)(\operatorname{snoc}(x, a)) & =f(\text { foldl }(c, f) x, a) .
\end{aligned}
$$

In standard functional programming, foldl ( $c, f$ ) is usually written in the form foldlf $c$ and is given as a function over the type listr $A$, since lists in functional programming are built with nil and cons rather than nil and snoc, but the above is an equivalent definition.

Now there is a standard result in the theory of lists, called Horner's rule, which says that if $f$ is monotonic with respect to $\leq$, then

$$
\max \cdot \operatorname{map}(\text { foldl }(c, f)) \cdot \text { tails }=\text { foldl }(c, g)
$$

where $g(a, b)=f(a, b) \sqcup c$ and $a \sqcup b$ is the greater of $a$ and $b$. We therefore obtain

$$
\text { max } \cdot \text { map sum } \cdot \text { tails }=\text { foldl }(0, \oplus)
$$

where $a \oplus b=(a+b) \sqcup 0$.
Using this result to continue the calculation of mss, again in the specific context of lists, we find

```
mss}=\operatorname{max}\cdotmap(foldl (0,\oplus))\cdot\mathrm{ inits,
```

where we have replaced subterms by inits. The final step is to make use of an important operation on lists called scanl, whose definition we will see in a moment. Scans are important in functional programming; in particular, Gibbons (1991) has made a study of scans on a particular species of binary tree. Over lists, the key fact is the Scan lemma, which says

$$
\operatorname{map}(\text { foldl }(c, f)) \cdot \text { inits }=\operatorname{scanl}(c, f)
$$

In this equation inits returns a list rather than a set, and map is a function on lists. In fact, inits returns the list of initial segments of a list in ascending order of length:

$$
\text { inits }\left[a_{1}, a_{2}, \ldots, a_{n}\right]=\left[[],\left[a_{1}\right],\left[a_{1}, a_{2}\right],\left[a_{1}, a_{2}, a_{3}\right], \ldots,\left[a_{1}, a_{2}, \ldots, a_{n}\right]\right]
$$

As a function returning lists, inits is defined by

$$
\begin{aligned}
\text { inits }= & \text { foldl }([n i l], f) \\
& \text { where } f(x, a)=\operatorname{snoc}(x, \text { snoc }(\text { last } x, a)) .
\end{aligned}
$$

The function scanl $(c, f)$ is defined similarly:

$$
\begin{aligned}
\operatorname{scanl}(c, f)= & \text { foldl }([c], g) \\
& \text { where } g(x, a)=\operatorname{snoc}(x, f(\text { last } x, a))
\end{aligned}
$$

Note that scanl (nil, snoc) = inits, so the definition of inits is a special case of scanl. The point of the scan lemma is that evaluation of

$$
\operatorname{scanl}(c, \oplus)\left[a_{1}, a_{2}, \ldots, a_{n}\right]=\left[c, c \oplus a_{1},\left(c \oplus a_{1}\right) \oplus a_{2}, \cdots\left(\left(c \oplus a_{1}\right) \oplus \cdots\right) \oplus a_{n}\right]
$$

can be done with $n$ evaluations of $\oplus$, whereas direct evaluation of map (foldl $(c, \oplus)$ ). inits requires $O\left(n^{2}\right)$ evaluations of $\oplus$ on a list of length $n$.

Applying the scan lemma to the problem of computing mss, we end up with the result that

$$
\begin{aligned}
m s s & =\max \cdot \operatorname{scanl}(0, \oplus) \\
a \oplus b & =(a+b) \sqcup 0
\end{aligned}
$$

where max is now interpreted as a function on lists rather than sets. The above identity leads at once to a linear time algorithm.

The rest of the paper is about how to generalise this calculation to arbitrary datatypes of the kind found in functional programming. In effect, we have to state and prove versions of Horner's rule and the Scan lemma that are valid for any datatype.

## 4 Datatypes

In what follows it is important to emphasise that, unless otherwise stated, a function means a total function whose source and target types are sets, unlike in standard functional programming where types are complete partial orders. As a departure from tradition, we reverse the usual order of writing the source and target types in function type declarations, preferring $f: A \leftarrow B$ rather than $f: B \rightarrow A$. This notation is consistent with adjectival order in English and has the advantage that the definition of function composition now takes the smooth form: if $f: A \leftarrow B$ and $g: B \leftarrow C$, then $f \cdot g: A \leftarrow C$.

We have already seen three examples of datatype declarations, namely those of listl $A$, listr $A$, and tree $A$. For example, let us recall

$$
\text { listr } A \quad::=\text { nil } \mid \text { cons }(A, \text { listr } A)
$$

Whenever one declares a datatype a number of functions are brought into play. In part, declaring a datatype as an equation asserts the existence of an isomorphism between the types on the left and right. In the case of listr $A$ this isomorphism takes the form

$$
\text { listr } A \cong 1+(A \times \operatorname{listr} A)
$$

The type 1 consists of just one member and serves as the source type for constants. It has the property that for each set $A$ there is precisely one function with type $1 \leftarrow A$, the usual notation for this function being $!_{A}$.

The type constructor $\times$ is cartesian product, and + is disjoint sum, also called coproduct. The right-hand side of the isomorphism can be rewritten, giving

$$
\operatorname{listr} A \cong \mathrm{~F}(A, \text { listr } A)
$$

where $\mathrm{F}(A, B)=1+(A \times B)$ is a mapping from types to types. We can also use F as a mapping from functions to functions by defining

$$
\mathrm{F}(f, g)=i d_{1}+(f \times g)
$$

The function $i d_{1}: 1 \leftarrow 1$ is the identity function on 1 . The cartesian product
constructor $\times$ is defined as a mapping between functions in the following way: if $f: A \leftarrow C$ and $g: B \leftarrow D$, then $f \times g: A \times B \leftarrow C \times D$ satisfies

$$
(f \times g)(c, d)=(f c, g d)
$$

Similarly, the coproduct constructor + can be defined on functions: applied to a left component $c$, the function $f+g: A+B \leftarrow C+D$ returns $f c$ as a left component of the result; dually, applied to a right component $d$, the value of $(f+g) d$ is the right component $g d$.

A function having a dual role both as a mapping between types and a mapping between functions is, provided certain properties are satisfied, called a functor. The functor $F$ defined above takes a pair of types or functions as argument and so is sometimes called a bifunctor. One property we require of a functor $F$ is that if $f: A \leftarrow B$, then $\mathrm{F} f: \mathrm{F} A \leftarrow \mathrm{~F} B$. The other properties are the identity and composition rules:

$$
\begin{aligned}
\mathrm{Fid}_{A} & =i d_{\mathrm{FA}_{A}} \\
\mathrm{~F}(f \cdot g) & =\mathrm{F} \cdot \mathrm{~F} g .
\end{aligned}
$$

The function $i d_{A}$ is the identity function with type $A \leftarrow A$. From now on we will usually omit the subscript on id, relying on context to resolve ambiguity.

In the case of bifunctors the above rules give, firstly, that if $f: A \leftarrow C$ and $g: B \leftarrow D$, then $\mathrm{F}(f, g): \mathrm{F}(A, B) \leftarrow \mathrm{F}(C, D)$; and, secondly, that

$$
\begin{aligned}
\mathrm{F}(i d, i d) & =i d \\
\mathrm{~F}(f \cdot \mathrm{~g}, h \cdot k) & =\mathrm{F}(f, h) \cdot \mathrm{F}(g, k)
\end{aligned}
$$

In particular, $x$ and + satisfy the identity and composition rules for bifunctors.
The functor $F$ associated with the declaration of a datatype is called the base functor of the declaration. Thus, $\mathrm{F}(A, B)=1+(A \times B)$ is the base functor associated with listr $A$. A functor is called polynomial if it is built up from constants, finite products and coproducts. More precisely,the class of polynomial functors is defined inductively by the following clauses:

1. The identity functor $i d$, defined by id $A=A$ and id $f=f$, and the constant functors $\mathrm{K}_{A}$, defined by $\mathrm{K}_{A} B=A$ and $\mathrm{K}_{A} f=i d_{A}$, are polynomial.
2. If $F$ and $G$ are polynomial, then so is their composition $F \cdot G$, their sum $F+G$ and their product $F \times G$, where

$$
\begin{aligned}
(\mathrm{F}+\mathrm{G}) f & =\mathrm{F} f+\mathrm{G} f \\
(\mathrm{~F} \times \mathrm{G}) f & =\mathrm{F} f \times \mathrm{G} f
\end{aligned}
$$

We will denote functors by single letters in sans serif font (because we need capital Roman letters for types and relations, introduced below), or by identifiers in ordinary italic font. In particular, id denotes both the identity function (on some given type) and also the identity functor.

The declaration of listr $A$ also introduces two functions

$$
\text { nil : listr } A \leftarrow 1 \quad \text { and } \quad \text { cons : listr } A \leftarrow A \times \operatorname{listr} A
$$

that serve to construct lists. We can parcel these functions together as one function

$$
[\text { nil, cons }]: \text { listr } A \leftarrow \mathrm{~F}(A, \text { listr } A) .
$$

In general, if $f: A \leftarrow B$ and $g: A \leftarrow C$, then $[f, g]: A \leftarrow B+C$ applies $f$ to left components and $g$ to right components. The function [nil, cons] has a special property, which captures the fact that we can define functions on lists by pattern-matching: given any function $[c, f]: B \leftarrow \mathrm{~F}(A, B)$ there is a unique function $h: B \leftarrow l i s t r A$ such that

$$
h \cdot[n i l, c o n s]=[c, f] \cdot F(i d, h)
$$

Unwrapping this compact equation, we get two equations

$$
\begin{aligned}
h \cdot n i l & =c \\
h \cdot \text { cons } & =f \cdot(i d \times h) .
\end{aligned}
$$

In functional programming $h$ is written in the form $h=$ foldr $(c, f)$, but we will use the alternative notation $h=([c, f])$. A function $h$ defined in this way is called a catamorphism, a term meaning 'according to form'. In functional programming, catamorphisms are what are known as fold operators. We have already met the fold operator corresponding to the type listl $A$, namely foldl.

Before describing the general situation, let us give one more example. The declaration of type tree $A$ in section 2 asserts tree $A \cong \mathrm{~F}(A$, tree $A)$, where this time the base functor $F$ is given by

$$
\mathrm{F}(A, B)=1+A \times B \times B
$$

The functor F is polynomial. The declaration of tree $A$ also introduces two functions

$$
\text { nil : tree } A \leftarrow 1 \quad \text { and } \quad \text { fork : } A \times \text { tree } A \times \text { tree } A
$$

that serve to construct trees. As before, we have

$$
[\text { nil, fork }]: \text { tree } A \leftarrow F(A \text {, tree } A) \text {. }
$$

The function [nil,fork] has a special property that given any function $[c, f]$ : $B \leftarrow \mathrm{~F}(A, B)$ there is a unique function $h: B \leftarrow$ tree $A$ such that

$$
h \cdot[\text { nil ,tree }]=[c, f] \cdot F(i d, h) .
$$

This time $h$ is a catamorphism on trees. Again we write $h=([c, f]$ ), so the notation (I - ] is implicitly parameterised by the base functor of the datatype.

Let us now consider the general situation. Think of the declaration of a datatype $\operatorname{term} A$ as providing two pieces of information: a polynomial base functor $\mathrm{F}(A, B)$, and a named function $\alpha: \operatorname{term} A \leftarrow F(A$, term $A$ ), which we will call the constructor of the type. The construtor $\alpha$ has the special property that given any function $f: B \leftarrow \mathrm{~F}(A, B)$ there is a unique function $h$, written $h=([f \mathrm{D})$, satisfying

$$
\begin{equation*}
h \cdot \alpha=f \cdot F(i d, h) \tag{1}
\end{equation*}
$$

As a consequence of this property of $\alpha$ we get that $[\alpha])=i d$. For example, ([nil, cons $]$ (which now we should write more accurately as $\mathbb{[}[$ nil , cons $] \mathbb{D}$ but won't) is the identity
function on lists. Less obviously, it also follows from its defining property that $\alpha$ is an isomorphism, meaning

$$
\alpha \cdot \alpha^{\circ}=i d \quad \text { and } \quad \alpha^{\circ} \cdot \alpha=i d
$$

where $\alpha^{\circ}$ denotes the inverse function to $\alpha$. The first id is the identity on term $A$, and the second is the identity on $\mathrm{F}(A$, term $A$ ), so $\alpha$ is the isomorphism that establishes $\operatorname{term} A \cong \mathrm{~F}(A$, term $A)$.

Using the fact that $h$ is uniquely characterised by (1), we obtain the following useful fusion rule for combining two functions into one:

$$
f \cdot[g])=[h]) \Leftarrow f \cdot g=h \cdot F(i d, f) .
$$

The proof is:

$$
\begin{aligned}
& f \cdot([g])=([h \rrbracket) \\
& \equiv \quad\{(1) \text { for }([h D\} \\
& f \cdot[g] \cdot \alpha=h \cdot F(i d, f \cdot[g]) \\
& \equiv \quad\{(1) \text { for }([g]\} \\
& f \cdot g \cdot \mathrm{~F}(i d,[g \mathrm{D})=h \cdot \mathrm{~F}(i d, f \cdot \llbracket g \mathrm{D}) \\
& \equiv \quad \text { \{property of functors\} } \\
& f \cdot g \cdot \mathrm{~F}(i d,[g \mathrm{D})=h \cdot \mathrm{~F}(i d, f) \cdot \mathrm{F}(i d,[g \mathrm{D}) \\
& \Leftarrow\} \\
& f \cdot g=h \cdot \mathrm{~F}(i d, f) .
\end{aligned}
$$

Since $\alpha$ is an isomorphism, we can move it to the other side of the defining equation for $([f])$. Thus, $h=\left(\int f D\right.$ is the unique solution of the equation

$$
h=f \cdot \mathrm{~F}(i d, h) \cdot \alpha^{\circ} .
$$

We will use this fact below when we generalise to relations.
One further function is introduced whenever we declare a datatype term $A$; this is a function $\operatorname{term} f$ with type $\operatorname{term} A \leftarrow \operatorname{term} B$ when $f: A \leftarrow B$. The definition is

$$
\operatorname{term} f=[\alpha \cdot F(f, i d)]
$$

In the case term $=$ listr this definition expands to the equations

$$
\begin{aligned}
\text { listr } f \cdot \text { nil } & =\text { nil } \\
\text { listr } f \cdot \text { cons } & =\text { cons } \cdot(f \times i d)
\end{aligned}
$$

and defines the familiar map operation on lists: listr $f x$ applies $f$ to every element of $x$. We denote the map operation on $\operatorname{term} A$ by term $f$ because term is a functor. It is immediate that term id $=i d$, and the proof that

$$
\operatorname{term}(f \cdot g)=\operatorname{term} f \cdot \operatorname{term} g
$$

is a simple exercise in the fusion law, which is left to the reader. We will call term
a type functor. Although it is not polynomial, we can allow it in the declarations of other datatypes without altering the theory given above. For instance, the type

$$
\text { tree } A \quad::=\text { node }(A, \text { listr }(\text { tree } A))
$$

introduces a datatype based on the non polynomial functor $\mathrm{F}(A, B)=A \times$ listr $B$. In this case the constructor function $\alpha=$ node.

We will call a datatype inductive if its base functor is polynomial or a type functor. In what follows we restrict attention to inductive datatypes.

## 5 Relations

Now, let us extend the foregoing theory to relations. We write $R: A \leftarrow B$ to denote that $R$ is a relation of type ' $A$ from $B$ '; we can think of $R$ as a subset of $A \times B$. Relational composition, like its functional counterpart, goes backwards: $R \cdot S$ is pronounced ' $R$ after $S$ '. We reserve single lower-case letters $f, g$ and so on, to denote functions.

Unlike functions, every relation has a converse. If $R: A \leftarrow B$ then the converse relation is $R^{\circ}: B \leftarrow A$. Converse preserves identities but reverses composition, so $(R \cdot S)^{\circ}=S^{\circ} \cdot R^{\circ}$.

For each $A$ and $B$ the relations of type $A \leftarrow B$ form a complete lattice with union $\cup$ and intersection $\cap$. Relations with the same type can be compared via a partial order $\subseteq$, where $R \subseteq S$ denotes $R \cap S=R$. We will sometimes use $\supseteq$ rather than $\subseteq$ in writing inequations because $\supseteq$ can be interpreted as refinement: $R \supseteq S$ if $R$ refines to $S$. Thus, a chain of inclusions

$$
S \supseteq S_{1} \supseteq \cdots \supseteq S_{n} \supseteq f
$$

can be interpreted as a stepwise refinement of a relational specification $S$ into a function (an executable program) $f$.

Converse preserves $\subseteq$ and composition distributes over (arbitrary) unions, but only weakly distributes over intersection in that

$$
R \cdot(S \cap T) \subseteq(R \cdot S) \cap(R \cdot T)
$$

We will suppose in what follows that composition binds more tightly than any other operation, so the right-hand side could have been written without brackets. Using the given properties of converse, we get from the above inequation a second one:

$$
(R \cap S) \cdot T \subseteq(R \cdot T) \cap(S \cdot T)
$$

These two inequations say that composition is monotonic in both arguments under $\subseteq$. One further inequation, called the modular law, is adjoined to the other axioms to give a weak converse of distributivity over intersection:

$$
(R \cdot S) \cap T \subseteq R \cdot\left(S \cap R^{\circ} \cdot T\right)
$$

There is more to be said about the calculus of relations, which is based on Freyd's theory of allegories (Freyd and Ščedrov, 1990), but we will postpone saying it to later. For now let us concentrate on the main point, which is that everything we have
said above about datatypes goes through when functions are extended to relations, provided only that we restrict attention to monotonic functors; that is, if $R \subseteq S$ then $F R \subseteq F S$. In particular, polynomial functors and type functors are monotonic. It can be shown that every functor on functions can be extended to a monotonic functor on relations in at most one way. It can also be shown that such functors preserve relational converse, so $(\mathrm{FR})^{\circ}=\mathrm{F}\left(R^{\circ}\right)$. It follows that the expression $\mathrm{F} R^{\circ}$ is not ambiguous.

By extending the theory to relations we get relational catamorphisms as well as functional ones. Moreover, since converse reverses composition we get, for a relation $R: \operatorname{term} A \leftarrow F(A$, term $A)$, not only that $X=(\mathbb{R}]$ is the unique solution of

$$
X=R \cdot F(i d, X) \cdot \alpha^{\circ}
$$

but also that $X=\left([R]^{\circ}\right.$ is the unique solution of

$$
X=\alpha \cdot F(i d, X) \cdot R^{\circ}
$$

By the Knaster-Tarski theorem the unique solution (if it exists) of $X=\phi X$ is also the least solution (under relational inclusion) of $X \supseteq \phi X$ and the greatest solution of $X \subseteq \phi X$, so we get all three of the following versions of the characterisation of catamorphisms:

$$
\begin{aligned}
X=R \cdot \mathrm{~F} X \cdot \alpha^{\circ} & \equiv X=([R] \\
X \subseteq R \cdot \mathrm{~F} X \cdot \alpha^{\circ} & \Rightarrow X \subseteq([R] \\
X \supseteq R \cdot \mathrm{~F} X \cdot \alpha^{\circ} & \Rightarrow X \supseteq(\mathbb{R})
\end{aligned}
$$

With relations we also get two variants of the fusion rule:

$$
\begin{aligned}
& R \cdot(\mathbb{S}) \subseteq \mathbb{I} T \mathbb{}) \Leftarrow R \cdot S \subseteq T \cdot F(i d, R) \\
& R \cdot(S D \supseteq \mathbb{C} \subseteq \Leftarrow R \cdot S \supseteq T \cdot F(i d, R) .
\end{aligned}
$$

Finally, writing ( $\mu X: \phi X$ ) for the least fixed point of $\phi$, we get the following rule:

$$
\begin{equation*}
\left[R \mathbb { ] } \cdot \left[S \mathbb{] ^ { \circ }}=\left(\mu X: R \cdot F(i d, X) \cdot S^{\circ}\right)\right.\right. \tag{2}
\end{equation*}
$$

With $S: C \leftarrow F(A, C)$ and $R: B \leftarrow \mathrm{~F}(A, B)$ we have $\left[[S]^{\circ}:\right.$ term $A \leftarrow C$ and $[R D: B \leftarrow \operatorname{term} A$, so (2) is a rule for eliminating the intermediate datatype term $A$ from a computation.

## 6 Prunings and Horner's rule

Let us now proceed to formalise the notion of pruning introduced in section 2, and to state and prove a general version of Horner's rule. For simplicity, we will assume that the base functor $F$ of the datatype term $A$ takes a particular form, namely

$$
\mathrm{F}(A, B)=1+\mathrm{G}(A, B)
$$

for some binary functor $G$ which is not further specified. We follow Backhouse (Aarts et al., 1992) in calling such functors pointed. Since $F(A, B)$ is a coproduct, the constructor function $\alpha:$ term $A \leftarrow F(A$, term $A)$ can be written in the form $\alpha=\left[\alpha_{0}, \alpha_{1}\right]$,
where $\alpha_{0}: \operatorname{term} A \leftarrow 1$ and $\alpha_{1}: \operatorname{term} A \leftarrow \mathrm{G}(A, \operatorname{term} A)$. One can think of $\alpha_{0}$ as a constant returning the empty element of $\operatorname{term} A$; in the sequel we will also use $\alpha_{0}$ as a constant function of type $\operatorname{term} A \leftarrow B$. More precisely, this constant function should be written as $\alpha_{0} \cdot!_{B}$, where $!_{B}$ is the unique function of type $1 \leftarrow B$. The same notational abbreviation will be used for other constants; thus, if $c: A \leftarrow 1$ we will also write $c$ for the constant function $c \cdot!_{B}: A \leftarrow B$.

Now, to prune a term $x$ means to substitute $\alpha_{0}$ for some (zero or more) subterms of $x$. The relation prune $: \operatorname{term} A \leftarrow \operatorname{term} A$ takes a term and prunes it in some arbitrary way:

$$
\text { prune }=\left[\left\{\alpha_{0}, \alpha_{0} \cup \alpha_{1}\right\rceil\right) .
$$

The first $\alpha_{0}$ has type term $A \leftarrow 1$, while the second has type term $A \leftarrow \mathrm{G}(A$, term $A)$.
The function prunings in Section 2 is the set of possible results returned by prune:

$$
\text { prunings } x=\{y \mid y \text { prune } x\} .
$$

In the relational calculus the mapping that associetes with each relation the corresponding set-valued function is denoted by $\Lambda$. Thus, if $R: A \leftarrow B$, then $\Lambda R: \mathrm{P} A \leftarrow B$, where $\mathrm{P} A$ denotes the powerset of $A$. In particular, prunings $=\Lambda$ prune. Writing P in sans serif font suggests that it is a functor; and indeed it is: Pf is the function that applies $f$ to every element of a set. Thus Pf is the function that we wrote as map $f$ in section 3.

So as not to lose the main thread in the mass of detail to come, we will now state the general version of Horner's rule, ignoring the fact that its formulation contains some concepts that we have not yet formally defined:

## Lemma 6.1

(Horner's rule). Let F be a pointed, monotonic functor and $f=\left[f_{0}, f_{1}\right]: B \leftarrow \mathrm{~F}(A, B)$ be a function. Furthermore, suppose $R: B \leftarrow B$ is a preorder such that $f$ is monotonic under $R$. Then

$$
\max R \cdot \mathrm{P}\left(\left[f \mathrm{D} \cdot \Lambda \text { prune } \supseteq\left(\max R \cdot \Lambda\left[f_{0}, f_{0} \cup f_{1}\right] \mathrm{D}\right)\right.\right.
$$

Horner's rule gives conditions under which one computation can be refined by another; the computation on the left takes the set of all prunings, applies a functional catamorphism to each pruning, and takes a maximum under a relation $R$; the computation on the right is a relational catamorphism that selects a maximum at each step.

Let us now explain the additional concepts in the statement of Horner's rule. First, a preorder is a reflexive, transitive relation, so $R: B \leftarrow B$ is a preorder if id $\subseteq R$ and $R \cdot R \subseteq R$. Next, and for the moment informally, $\max R: B \leftarrow \mathrm{~PB}$ is a relation that, given a set, returns some maximum element under $R$. Finally, a function $f: B \leftarrow F(A, B)$ is monotonic under $R$ if

$$
f \cdot \mathrm{~F}(i d, R) \subseteq R \cdot f
$$

Here are two examples to explain the definition of monotonicity:

1. Addition. Let $R=$ leq, where leq $: N a t \leftarrow$ Nat denotes the relation $\leq$ on natural numbers, and let $f=$ plus, where plus : Nat $\leftarrow$ Nat $\times$ Nat denotes binary addition. Taking $\mathrm{F}(A, B)=A \times B$ the monotonicity condition translates to

$$
p l u s \cdot(i d \times l e q) \subseteq l e q \cdot p l u s
$$

and says that $x=y+z$ and $z \leq z^{\prime}$ implies $x \leq y+z^{\prime}$. This is just the (true) statement that addition is monotonic in its right argument. We can also take $\mathrm{F}(A, B)=B \times B$, in which case the monotonicity condition is

$$
p l u s \cdot(l e q \times l e q) \subseteq l e q \cdot p l u s
$$

and asserts that addition is monotonic in both arguments.
2. Cons lists. Let $R=$ lex, where lex is the lexicographic ordering on lists, and $f=[$ nil, cons $]$. With $\mathrm{F}(A, B)=1+A \times B$ the monotonicity condition translates to

$$
\begin{aligned}
n i l & \subseteq l e x \cdot n i l \\
\text { cons } \cdot(i d \times l e x) & \subseteq l e x \cdot c o n s
\end{aligned}
$$

The first equation follows at once from the reflexivity of lex, and the second asserts the true statment that cons is monotonic with respect to the lexicographic ordering.

To prove Horner's rule we need to define $\max R$ formally in the relational calculus, and also to be more precise about the relationship between $P$ and $\Lambda$. We begin with the latter.

### 6.1 Powersets

Formally, the isomorphism between relations and set-valued functions can be described in the following suitably abstract form. For every set $A$ there exists a set $\mathrm{P} A$, called the powerset of $A$, and a relation $\in: A \leftarrow \mathrm{P} A$, called the membership relation on $A$, which together are characterised by the following property: for every relation $R: A \leftarrow B$, there exists a function $\Lambda R: \mathrm{P} A \leftarrow B$ such that

$$
(f=\Lambda R) \equiv(\in \cdot f=R) \quad \text { for all } f: \mathrm{P} A \leftarrow B
$$

The function $\Lambda R$ is said to be the power transpose of $R$ and can be defined in set theory by $(\Lambda R) b=\{a \mid a R b\}$. (In fact, much of set theory can be recovered using just this universal property of powersets, plus the relational calculus. This observation lies at the heart of the categorical approach to sets, namely the theory of toposes (Johnstone, 1977; Barr and Wells, 1985; Goldblatt, 1986).)

It is immediate from the universal property of $\Lambda$ that

$$
\epsilon \cdot \Lambda R=R
$$

Below we refer to this fact by the hint ' $\Lambda$ cancellation'. It also follows from the universal property that $i d: \mathrm{P} A \leftarrow \mathrm{P} A$ satisfies

$$
i d=\Lambda(\in)
$$

Using $\Lambda$ we can define the existential image of a relation $R: A \leftarrow B$; this is a function $E R: P A \leftarrow P B$ defined by

$$
E R=\Lambda(R \cdot \epsilon)
$$

In set theory we have

$$
(\mathrm{ER}) x=\{a \mid(\exists b: a R b \wedge b \in x)\}
$$

It is clear from its definition that Eid $=i d$, and below we will show that $E(R \cdot S)=$ $E R \cdot E S$, so $E$ is a functor (taking its action on types to be $E A=P A$ ). It is not, however, a monotonic functor on relations because it returns a function and $f \subseteq g$ if and only if $f=g$.

To show E is a functor we first prove $\Lambda(R \cdot S)=\mathrm{E} R \cdot \Lambda S$ :

$$
\begin{array}{lc} 
& \Lambda(R \cdot S)=\mathrm{E} R \cdot \Lambda S \\
\equiv & \{\text { definition of } \boldsymbol{\Lambda}\} \\
& R \cdot S=\epsilon \cdot \mathrm{E} R \cdot \Lambda S \\
\equiv & \{\text { definition of } \mathrm{E}\} \\
& R \cdot S=\epsilon \cdot \Lambda(R \cdot \epsilon) \cdot \Lambda S \\
\equiv & \{\Lambda \text { cancellation (twice })\} \\
& \text { true }
\end{array}
$$

Now, taking $S=T \cdot \in$, we get $\mathrm{E}(R \cdot T)=\mathrm{E} R \cdot \mathrm{E} T$.
Although $E$ is not monotonic, there does exist a variant of $E$ which is, namely the powerset functor $P$. Over functions, this functor has the same action as $E$, so $\mathrm{Pf}=\mathrm{E} f$. For a relation $R$ the definition of PR in set theory is

$$
x \mathrm{PR} y=(\forall a \in x: \exists b \in y: a R b) \wedge(\forall b \in y: \exists a \in x: a R b)
$$

For use below, we note the following two identities, in which the function union : $\mathrm{P} A \leftarrow \mathrm{PP} A$ is defined by union $=\mathrm{E}(\epsilon)$; this function returns the union of a collection of sets. The identities are:

$$
\begin{align*}
\Lambda(R \cdot S) & =\text { union } \cdot \mathrm{P} \Lambda R \cdot \Lambda S  \tag{3}\\
\mathrm{Pf} \cdot \text { union } & =\text { union } \cdot \mathrm{PP} f . \tag{4}
\end{align*}
$$

Equation (4) appeared in section 3. Both equations are easy consequences of the definitions above and we omit details; for a further discussion of $P$ and its relation to E, see De Moor (1992).

### 6.2 Division

To define max $R$ we need the operation of relational division. Because relational composition distributes over arbitrary unions, it has a weak inverse, called division, which is characterised by the equivalence

$$
T \subseteq R / S \equiv T \cdot S \subseteq R \quad \text { for all } T
$$

The operator / can be defined in set theory by

$$
a(R / S) b \equiv(\forall c: b S c: a R c)
$$

A second division operator $\backslash$ can be introduced by defining $R \backslash S=\left(S^{\circ} / R^{\circ}\right)^{\circ}$, so

$$
T \subseteq R \backslash S \equiv R \cdot T \subseteq S \quad \text { for all } T
$$

As a predicate we have $a(R \backslash S) b \equiv(\forall c: c R a: c S b)$.
Above we defined $P R$ in set theoretical terms; using division we can define

$$
\mathrm{P} R=\epsilon \backslash(R \cdot \epsilon) \cap(\ni \cdot R) / \ni
$$

where $\ni$ is shorthand for $\epsilon^{\circ}$. The expression on the right is the translation of the earlier one into the relational calculus.

Using division we can define the relation $\max R: A \leftarrow \mathrm{P} A$ by

$$
\max R=\in \cap\left(R^{\circ} / \ni\right)
$$

This definition corresponds to the usual one in set theory: $a(\max R) x$ holds when $a$ is an element of $x$ (the first term) and $x$ has upper bound $a$ (the second term), that is, for all $b \in x$, we have $b R a$. Although the definition of $\max R$ does not require $R$ to be a preorder, it is useful only when $R$ is one, so we will tacitly assume it so.

### 6.3 Properties of max

There are two properties of max that we will need. First of all,

$$
\begin{equation*}
X \subseteq \max R \cdot \Lambda S \equiv(X \subseteq S) \wedge\left(X \cdot S^{\circ} \subseteq R^{\circ}\right) \tag{5}
\end{equation*}
$$

We give the proof of (5) because it is typical of the kind of manipulations found in the relational calculus. The calculation makes use of the following two rules in whuch $f$ is a function and $R$ and $S$ arbitrary relations. Firstly, we have the distributive law

$$
(R \cap S) \cdot f=(R \cdot f) \cap(S \cdot f)
$$

and secondly the shunting law

$$
R \subseteq S \cdot f \equiv R \cdot f^{\circ} \subseteq S
$$

The proof of (5) is:

$$
\begin{aligned}
& X \subseteq \max R \cdot \Lambda S \\
\equiv & \{\text { definition of } \max R\} \\
& X \subseteq\left(\in \cap\left(R^{\circ} / \ni\right)\right) \cdot \Lambda S \\
\equiv & \{\text { distributive law (above })\} \\
& X \subseteq(\in \cdot \Lambda S) \cap\left(\left(R^{\circ} / \ni\right) \cdot \Lambda S\right) \\
\equiv & \{\Lambda \text { cancellation and universal property of } \cap\} \\
& (X \subseteq S) \wedge\left(X \subseteq\left(R^{\circ} / \ni\right) \cdot \Lambda S\right)
\end{aligned}
$$

Continuing with the second term:

$$
\begin{array}{ll} 
& X \subseteq\left(R^{\circ} / \ni\right) \cdot \Lambda S \\
\equiv & \{\text { shunting law (above) }\} \\
& X \cdot(\Lambda S)^{\circ} \subseteq R^{\circ} / \ni \\
\equiv & \quad \text { universal property of } /\} \\
& X \cdot(\Lambda S)^{\circ} \cdot \ni \subseteq R^{\circ} \\
\equiv & \{\text { converse and } \Lambda \text { cancellation }\} \\
& X \cdot S^{\circ} \subseteq R^{\circ} .
\end{array}
$$

The second fact is that max $R$ weakly distributes over union:

$$
\begin{equation*}
\max R \cdot \text { union } \supseteq \max R \cdot P(\max R) \tag{6}
\end{equation*}
$$

We will sketch the proof since it uses (5). Since union $=E(\epsilon)=\Lambda(\epsilon \cdot \epsilon)$, condition (5) shows it is sufficient to prove:

$$
\begin{aligned}
\max R \cdot \mathrm{P}(\max R) & \subseteq \epsilon \cdot \epsilon \\
\max R \cdot \mathrm{P}(\max R) \cdot \ni \cdot \ni & \subseteq R^{\circ}
\end{aligned}
$$

The first is easy using $\max R \subseteq \epsilon$ and $\in \cdot \mathrm{PX} \subseteq X \cdot \epsilon$, and for the second we use $\mathrm{P} X \cdot \ni \subseteq \ni \cdot X$ and max $R \cdot \ni \subseteq R^{\circ}$ to write the inclusion in the form $R^{\circ} \cdot R^{\circ} \subseteq R^{\circ}$, which follows from the transitivity of $R$.

We saw a version of (6) in section 3 where it appeared as an equation, but with the qualification that it applied only to sets of non-empty sets. It is one of the advantages of a relational approach that we can omit the qualification provided we replace equality by refinement.

### 6.4 Proof of Horner's rule

We are now ready for the proof of Horner's rule. Our aim is to apply the universal property for $\max R$, so we begin

$$
\begin{aligned}
& \max R \cdot \mathrm{P}(f f] \cdot \Lambda \text { prune } \\
= & \{\text { since } \mathrm{P}=\mathrm{E} \text { on functions }\} \\
& \max R \cdot \mathrm{E}(f \mathrm{f}] \cdot \Lambda \text { prune } \\
= & \{\text { since } \mathrm{E} X \cdot \Lambda Y=\Lambda(X \cdot Y)\} \\
& \max R \cdot \Lambda(\mathbb{U} \mathrm{D} \cdot \text { prune }) \\
= & \{\text { definition of prune }\} \\
& \max R \cdot \Lambda\left(( \mathbb { J } ) \cdot \left(\left[\alpha_{0}, \alpha_{0} \cup \alpha_{1} \mathbb{D}\right)\right.\right. \\
=\quad & \{\text { fusion }(\text { see below })\} \\
& \max R \cdot \Lambda\left(f_{0}, f_{0} \cup f_{1} \mathrm{D} .\right.
\end{aligned}
$$

The condition for fusion is that

$$
(f f) \cdot\left[\alpha_{0}, \alpha_{0} \cup \alpha_{1}\right]=\left[f_{0}, f_{0} \cup f_{1}\right] \cdot F(i d,(f f)
$$

and is easily verified.
Abbreviating $\left[f_{0}, f_{0} \cup f_{1}\right]$ by $S$ and appealing to (5), we get Horner's rule by showing

$$
\begin{aligned}
(\max R \cdot \Lambda S] & \subseteq[S] \\
([\max R \cdot \Lambda S]) \cdot\left([S]^{\circ}\right. & \subseteq R^{\circ} .
\end{aligned}
$$

The first inequation is easy, since $\max R \cdot \Lambda S \subseteq \in \cdot \Lambda S=S$ and catamorphisms are monotonic. For the second inequation we appeal to (2) and show

$$
\max R \cdot \Lambda S \cdot \mathrm{~F}\left(i d, R^{\circ}\right) \cdot S^{\circ} \subseteq R^{\circ}
$$

To do this, we need the monotonicity of $f$ under $R$. It is easy to show that this condition implies that $S$ is also monotonic under $R$ in the sense that

$$
S \cdot F(i d, R) \subseteq R \cdot S
$$

Taking converses, and recalling the assumption that $F$ is a monotonic functor and so preserves converse, we obtain

$$
\mathrm{F}\left(i d, R^{\circ}\right) \cdot S^{\circ} \subseteq S^{\circ} \cdot R^{\circ}
$$

Now we can argue:

$$
\begin{array}{ll} 
& \max R \cdot \Lambda S \cdot \mathrm{~F}\left(i d, R^{\circ}\right) \cdot S^{\circ} \\
\subseteq & \{\text { above }\} \\
& \max R \cdot \Lambda S \cdot S^{\circ} \cdot R^{\circ} \\
\subseteq & \left\{\text { since } \Lambda X \cdot X^{\circ} \subseteq \ni \text { by shunting and } \Lambda \text { cancellation }\right\} \\
& \max R \cdot \ni \cdot R^{\circ} \\
\subseteq & \left\{\text { since } \max R \subseteq R^{\circ} / \ni \text { and universal property of } /\right\} \\
& R^{\circ} \cdot R^{\circ} \\
\subseteq & \{\text { converse and transitivity of } R\} \\
& R^{\circ} .
\end{array}
$$

The proof of Horner's rule is complete.
Before we can solve the problem of defining subterms and generalising the Scan lemma, we need to give some more theory about datatypes.

## 7 Natural transformations and membership

Datatypes record the presence of elements, so one would expect a type term $A$ to come equipped with a membership relation $\delta_{A}: A \leftarrow \operatorname{term} A$ such that $a \delta_{A} x$ precisely when $a$ is an element of $x$. Note that $\delta_{A}$ is a collection of relations, one for each type $A$. We have already seen one membership relation, namely $\epsilon_{A}: A \leftarrow \mathrm{P} A$, the ordinary membership relation for sets. To define the notion of membership for an arbitrary datatype, we first explain what it means for a collection of relations to be a natural transformation.

### 7.1 Natural transformations

A number of functions that we have met already are really collections of functions; for example

$$
\begin{array}{r}
i d_{A}: A \leftarrow A \\
\alpha_{A}: \text { term } A \leftarrow \mathrm{~F}(A, \text { term } A) \\
{\text { pruning } s_{A}}: \mathrm{P}(\text { term } A) \leftarrow \operatorname{term} A .
\end{array}
$$

The functions in each collection do not depend in any essential way on the parameter $A$, a fact which is captured by a suitable 'type changing' rule. For example, for any function $f: A \leftarrow B$ we have

$$
\begin{aligned}
f \cdot i d_{B} & =i d_{A} \cdot f \\
\operatorname{term} f \cdot \alpha_{B} & =\alpha_{A} \cdot F(f, \text { term } f)
\end{aligned}
$$

The second identity comes from the definition $\operatorname{term} f=[\alpha \cdot F(f, i d)]$. Formally, a collection of functions $\phi_{A}: \mathrm{FA} \leftarrow \mathrm{G} A$ is called a natural transformation if for all $f: A \leftarrow B$ we have

$$
\mathrm{F} f \cdot \phi_{B}=\phi_{A} \cdot \mathrm{G} f
$$

We write $\phi: \mathrm{F} \leftarrow \mathrm{G}$ to indicate that $\phi$ is natural. In particular, we have id :id $\leftarrow i d$, and $\alpha:$ term $\leftarrow \mathrm{G}$, where $\mathrm{G} A=\mathrm{F}(A$, term $A)$.

The notion of natural transformation extends to relations: a collection of relations $\phi: \mathrm{F} \leftarrow \mathrm{G}$ is natural if for all relations $R: A \leftarrow B$, we have $\mathrm{F} R \cdot \phi_{B}=\phi_{A} \cdot \mathrm{G} R$. For example, we have $\in: i d \leftarrow E$.

There is a weaker notion of natural transformation, more useful when dealing with relations: a collection of relations $\phi_{A}: \mathrm{F} A \leftarrow \mathrm{G} A$ is a weak natural transformation when for all $R: A \leftarrow B$ we have

$$
\mathrm{F} R \cdot \phi_{B} \supseteq \phi_{A} \cdot \mathrm{G} R .
$$

We write $\phi: \mathrm{F} \hookleftarrow \mathrm{G}$ to indicate that $\phi$ is weakly natural. It is a fact that $\phi$ is weakly natural if and only if $\mathrm{F} f \cdot \phi_{B}=\phi_{A} \cdot \mathrm{G} f$ for all functions $f: A \leftarrow B$. For a proof see Carboni et al. (1991). Thus, we can show $\phi$ is weakly natural for relations by showing it is natural for functions.

An important example is $\in:$ id $\hookleftarrow \mathrm{P}$; we have $R \cdot \in \supseteq \in \cdot \mathrm{P} R$ for all $R$, but this inclusion cannot be strengthened to an equality.

### 7.2 Membership

Now, let us return to membership. Formally, a collection of arrows $\delta_{A}: A \leftarrow \mathrm{~F} A$ is a membership relation for F if for each $R: A \leftarrow B$

$$
\mathrm{F} R \cdot\left(\delta_{B} \backslash i d_{B}\right)=\delta_{A} \backslash R
$$

Rather than attempt to explain this definition, we will give a number of properties that justify it. For formal proofs of these properties, see De Moor (1993). First, the
above equation has at most one solution for $\delta$ because any $\delta$ satisfying it is the largest weak natural transformation of type id $\hookleftarrow \mathrm{F}$.

Second, although not every functor has membership, all polynomial functors do. These membership relations are given inductively by the following clauses:

$$
\begin{aligned}
\delta_{\mathrm{id}} & =\text { id } \\
\delta_{\mathrm{K} A} & =\emptyset \\
\delta_{\mathrm{F}+\mathrm{G}} & =\left[\delta_{\mathrm{F}}, \delta_{\mathrm{G}}\right] \\
\delta_{\mathrm{F} \times \mathrm{G}} & =\delta_{\mathrm{F}} \cdot \text { outl } \cup \delta_{\mathrm{G}} \cdot \text { outr } \\
\delta_{\mathrm{F} \cdot \mathrm{G}} & =\delta_{\mathrm{G}} \cdot \delta_{\mathrm{F}}
\end{aligned}
$$

In the second clause $\emptyset$ denotes the empty relation; in the fourth clause the functions outl : $A \leftarrow A \times B$ and outr $: B \leftarrow A \times B$ are the projection functions. These functions have the property that for every two functions $f: A \leftarrow C$ and $g: B \leftarrow C$ there is a unique function $\langle f, g\rangle: A \times B \leftarrow C$ such that outl $\cdot\langle f, g\rangle=f$ and outr $\cdot\langle f, g\rangle=g$. Since we can define $f \times g=\langle f \cdot$ outl, $g \cdot$ outr $\rangle$ we get that outl and outr satisfy

$$
\begin{aligned}
& f \cdot \text { outl }=\text { outl } \cdot(f \times g) \\
& g \cdot \text { outr }=\text { outr } \cdot(f \times g),
\end{aligned}
$$

and so are natural transformations over functions. However, these equations are weakened when we consider relations:

$$
\begin{aligned}
& R \cdot \text { outl } \supseteq \text { outl } \cdot(R \times S) \\
& S \cdot \text { outr } \supseteq \text { outr } \cdot(R \times S) .
\end{aligned}
$$

These inclusions cannot be strengthened because, for instance, $R \times \emptyset=\emptyset$.
For type functors the question of membership is more complicated. Recall that term $f$ was defined as a catamorphism, so one might expect that its membership relation can also be expressed as a catamorphism. However, this is only true for datatypes that do not contain constants. For example, consider the type

$$
\operatorname{listr}^{+} A::=\quad \text { wrap } A \mid \operatorname{cons}^{+}\left(A, \text { listr }^{+} A\right)
$$

of non-empty lists, with base functor $\mathrm{F}(A, B)=A+A \times B$. The membership relation $\delta$ for these lists is given by

$$
\delta=([\text { id }, \text { outl } \cup \text { outr }\rceil)
$$

In words, an arbitrary member is obtained at each stage either by selecting the new element or by retaining the chosen element from the previous stage.

With the type listr $A$, in which the constructor wrap is replaced by nil, the membership relation is not given by

$$
\delta=\llbracket \emptyset, \text { outl } \cup \text { outr } \mathbb{D}
$$

because the right-hand side is the empty relation $\emptyset$, as one can easily check.
Fortunately, there is another approach to membership of type functors, one that makes use of two auxiliary relations, which we will call root and spur. To fix
notation, let $\alpha: \operatorname{term} A \leftarrow \mathrm{~F}(A, \operatorname{term} A)$ be the constructor of $\operatorname{term} A$ and let

$$
\begin{array}{r}
\text { left }: A \leftarrow \mathrm{~F}(A, B) \\
\text { right }: B \leftarrow \mathrm{~F}(A, B)
\end{array}
$$

be the two membership relations associated with the binary functor $F$. More precisely, left is the membership relation for the functor $\mathrm{G}_{B}(A)=\mathrm{F}(A, B)$ in which $B$ is fixed; similarly for right. The relations root $: A \leftarrow \operatorname{term} A$ and spur $: \operatorname{term} A \leftarrow \operatorname{term} A$ are defined by

$$
\begin{aligned}
\text { root } & =\text { left } \cdot \alpha^{\circ} \\
\text { spur } & =\text { right } \cdot \alpha^{\circ} .
\end{aligned}
$$

Let us explain these relations with the help of some examples.

1. Snoc lists. With $\mathrm{F}(A, B)=1+(B \times A)$, we get left $=[\emptyset$, outr $]$ and right $=$ [ $\emptyset$, outl], so

$$
\text { root }=\text { left } \cdot \alpha^{\circ}=[\emptyset, \text { outr }] \cdot[\text { nil }, \text { snoc }]^{\circ}=\text { outr } \cdot \text { snoc }^{\circ} .
$$

This uses the law $[R, S] \cdot[T, U]^{\circ}=R \cdot T^{\circ} \cup S \cdot U^{\circ}$. Similarly, we get spur $=$ outl $\cdot$ snoc $^{\circ}$. In words, root is last, the partial function that returns the last element of a (nonempty) list, and spur is init, the partial function that removes the last element.
2. Cons lists. Dually, with the base functor $\mathrm{F}(A, B)=1+(A \times B)$, we get root $=$ outl $\cdot$ cons $^{\circ}$ and spur $=$ outr $\cdot$ cons $^{\circ}$, so root is the partial function that returns the first element of a list and spur removes it.
3. Binary trees. With $\mathrm{F}(A, B)=1+(A \times(B \times B))$, we get the type tree $A$ described in section 2. Here we have

$$
\begin{aligned}
\text { left } & =[\emptyset, \text { outl }] \\
\text { right } & =[\emptyset,(\text { outl } \cup \text { outr }) \cdot \text { outr }]
\end{aligned}
$$

so

$$
\begin{aligned}
\text { root } & =\text { outl } \cdot \text { fork }^{\circ} \\
\text { spur } & =(\text { outl } \cup \text { outr }) \cdot \text { outr } \cdot \text { fork }^{\circ} .
\end{aligned}
$$

The partial function root returns the root of a nonempty tree, and spur returns one of its immediate subtrees.

Now we can define the membership relation $\delta: A \leftarrow \operatorname{term} A$ by

$$
\delta=\text { root } \cdot \text { spur }^{*}
$$

where $R^{*}$ denotes the reflexive transitive closure of $R$. In the case of snoc-lists, an arbitrary member of a list is obtained by taking the last element of an arbitrary prefix; in the case of cons-lists, an arbitrary member is obtained by taking the first element of an arbirtrary suffix; and in the case of trees, an arbitrary member is obtained by taking the root of an arbitrary subtree.

The last sentence suggests that we now have a way to define an arbitrary subterm
of a term: define subterm $=$ spur** The set of subterms is then obtained as $\Lambda$ subterm. This is perfectly correct, except that in the next section we will define subterms not as a set of terms but as a certain datatype containing terms as elements. The reason for this lies in the formulation of the Scan lemma, which depends critically in the case of lists on subterms returning a list of lists.

Since we will be using such datatypes to represent sets, we will need a way of moving from one to the other. The function setify : PA term $A$ takes an element of the datatype $\operatorname{term} A$ and returns a set of its elements. The definition is very simple: setify $=\Lambda \delta$, where $\delta$ is the membership relation for term. The function setify is a weak natural transformation setify : P $\hookleftarrow$ term, so

$$
\begin{equation*}
\mathrm{PR} \cdot \text { setify } \quad \supseteq \text { setify } \cdot \text { term } R . \tag{7}
\end{equation*}
$$

The proof, which we omit, uses the formal definition of membership for a datatype.
Finally, we will need to know how to implement the relation $\max R$ on datatypes other than sets. With left and right as above we have

$$
\begin{equation*}
\max R \cdot \text { setify } \quad \supseteq \quad(\max R \cdot \Lambda(\text { left } \cup \text { right }) \mathrm{D} . \tag{8}
\end{equation*}
$$

Note that in the catamorphism on the right the relations left and right both have type $A \leftarrow \mathrm{~F}(A, A)$. Note also that the inclusion cannot be strengthened to an equality because there may be constant terms that have no elements.

## 8 Subterms and scans

Recall from section 2 that we defined subterms as a set-valued function that returned all possible subterms of a given struture. In analogy with the definition of prunings it is therefore tempting to define subterms $=\Lambda$ subterm, where subterm was defined above. But also recall the Scan lemma in section 3 which depends on inits $=$ subterms returning a list of subterms.

In an attempt to construct a generic version of the scan lemma, one might think of generating subterms as a list of structures, but that would still involve lists in an essential way. We really want to think of a structure of structures: a list of lists, a tree of trees, and so on. The way to achieve this is to create a new type of labelled structures in which each 'node' is labelled with the corresponding subterm. Not every datatype allows the labelling of nodes (think of unlabelled binary trees), but there is a canonical way of introducing labels into a datatype. We consider this first, returning to scans at the end of the section.

### 8.1 Labelled datatypes

Again, let $\alpha: \operatorname{term} A \leftarrow \mathrm{~F}(A$, term $A)$ be the constructor of term $A$. Define another bifunctor $\mathrm{F}^{+}$by

$$
\mathrm{F}^{+}(A, B)=\mathrm{F}(1, B) \times A
$$

and let this be the base functor of a datatype $\operatorname{term}^{+} A$ of 'labelled terms'. Let $\alpha^{+}:$term $^{+} A \leftarrow \mathrm{~F}^{+}\left(A\right.$, term $\left.^{+} A\right)$ be its constructor. Here are some examples to clarify the idea.

1. Cons lists. With $\mathrm{F}(A, B)=1+(A \times B)$ we get

$$
\mathrm{F}^{+}(A, B)=(1+(1 \times B)) \times A \cong A+(A \times B)
$$

So labelled cons-lists are isomorphic to the type listr ${ }^{+}$of non-empty cons lists.
2. Non-empty cons lists. What happens when we try and label non-empty cons lists? Here $\mathrm{F}(A, B)=A+(A \times B)$ and so

$$
\mathrm{F}^{+}(A, B)=(1+(1 \times B)) \times A \cong A+(A \times B)
$$

so labelling does not change non-empty cons lists in any essential way.
3. Binary trees. With $\mathrm{F}(A, B)=1+(A \times B \times B)$, we find

$$
\mathrm{F}^{+}(A, B)=(1+(1 \times B \times B)) \times A \cong A+(A \times B \times B)
$$

Labelled trees are therefore isomorphic to the type

$$
\text { tree }^{+} A::=\text { tip } A \mid \text { fork }^{+}\left(A, \text { tree }^{+} A, \text { tree }^{+} A\right)
$$

of non-empty labelled trees.

### 8.2 Subterms

We can now define subterms $:$ term $^{+}(\operatorname{term} A) \leftarrow \operatorname{term} A$ as the catamorphism

$$
\begin{equation*}
\text { subterms }=(\lfloor a c c \alpha\rfloor, \tag{9}
\end{equation*}
$$

where for $R: B \leftarrow \mathrm{~F}(A, B)$ the relation acc $R$ (short for 'accumulate with $R$ ') has type

$$
\text { acc } R: \text { term }^{+} B \leftarrow \mathcal{F}\left(A, \text { term }^{+} B\right)
$$

and is given by the complicated expression

$$
\text { acc } R=\alpha^{+} \cdot\left\langle\mathrm{F}(!, i d), R \cdot \mathrm{~F}\left(i d, \text { root }^{+}\right)\right\rangle,
$$

and where root ${ }^{+}: \operatorname{term} A \leftarrow$ term $^{+}($term $A)$ is the (total!) function root $=$ outl $\cdot\left(\alpha^{+}\right)^{\circ}$. The definition is somewhat opaque, so we will give some examples.

1. Snoc lists. We have just seen that labelled snoc lists are isomorphic to the type

$$
\text { listl }^{+} A \quad::=\quad \operatorname{wrap} A \mid \operatorname{snoc}^{+}\left(\text {listl }^{+} A, A\right)
$$

of non-empty snoc-lists. Here, root ${ }^{+}=$last, the (total!) function that returns the last element of a list of type listl ${ }^{+} A$. Let $\sigma$ denote the isomorphism

$$
\sigma: A+(B \times A) \leftarrow(1+(1 \times B)) \times A .
$$

Now we have

$$
\begin{aligned}
& \text { acc } \alpha \\
= & \{\text { definitions, and } \mathrm{F}(A, B)=1+(B \times A)\} \\
& \quad\left[\text { wrap }, \text { snoc }{ }^{+}\right] \cdot \sigma \cdot\langle\text { id }+(\text { id } \times!),[\text { nil }, \text { snoc }] \cdot(\text { id }+ \text { last } \times \text { id })\rangle \\
= & \{\text { coproduct law }\}
\end{aligned}
$$

$$
\left.\begin{array}{rl} 
& {\left[\text { wrap, snoc }{ }^{+}\right] \cdot \sigma \cdot\langle\text { id }+(\text { id } \times!),[\text { nil }, \text { snoc } \cdot(\text { last } \times \text { id })]\rangle} \\
= & \{\text { property of } \sigma\}
\end{array}\right\}
$$

The coproduct law used above is that $[R, S] \cdot(T+U)=[R \cdot T, S \cdot U]$. Writing [nil] for the constant returned by wrap $\cdot$ nil and snoc for snoc ${ }^{+}$(thereby embedding the type $l i s t l^{+} A$ in listl $A$ ), we get

$$
\begin{aligned}
\text { subterms } & =\text { foldl }([\text { nil }], f) \\
f(x s, a) & =\operatorname{snoc}(x s, \text { snoc }(\text { last xs }, a))
\end{aligned}
$$

This is precisely the definition of the function inits we met in section 3.
2. Binary trees. Here labelled trees are isomorphic to the type

$$
\operatorname{tree}^{+} A::=\text { tip } A \mid \text { fork }^{+}\left(A, \text { tree }^{+} A, \text { tree }^{+} A\right)
$$

of non-empty labelled trees. This time root ${ }^{+}$is the total function that returns the element at the root of the tree. As before, we can calculate

$$
\operatorname{acc} \alpha=\left[t i p \cdot n i l, \text { fork }{ }^{+} \cdot\left\langle\text { id } \times \text { id, fork } \cdot\left(\text { id } \times \operatorname{root}^{+} \times \text {root }^{+}\right)\right\rangle\right] .
$$

As in the case of snoc-lists we can embed tree ${ }^{+} A$ in tree $A$ by taking tip $a=$ fork ( $a$, nil, nil). The result is a computation for subterms that a functional programmer would write in the form

$$
\begin{aligned}
\text { subterms } & =\text { foldtree }(c, f) \\
c & =\text { fork }(\text { nil, nil, nil }) \\
f(a, x, y) & =\text { fork }(\text { fork }(a, \text { root } x, \text { root } y), x, y) \\
\text { root }(\text { fork }(a, x, y)) & =a \\
\text { foldtree }(c, f) \text { nil } & =c \\
\text { foldtree }(c, f)(\text { fork }(a, x, y)) & =f(a, \text { foldtree }(c, f) x, \text { foldtree }(c, f) y)
\end{aligned}
$$

### 8.3 Properties of subterms

Let us now look at two properties of subterms. First, subterms is a function because only functions appear in its definition (recall that root ${ }^{+}=$outl $\cdot\left(\alpha^{+}\right)^{\circ}$ is a total function).

The second, more important fact is that

$$
\begin{equation*}
\text { setify } \cdot \text { subterms }=\text { ^subterm }, \tag{10}
\end{equation*}
$$

where subterm $=s p u r^{*}$. The proof, which we omit, depends on the fact that spur* is
the unique solution for $X$ of the equation

$$
X=i d \cup(X \cdot s p u r)
$$

### 8.4 Scans and the scan lemma

Now let us return to scans and the scan lemma. We saw in section 3 that inits $=$ scanl (nil, snoc). The same is true of the general scan, which is simply (acc R]). The scan lemma therefore takes the form

## Lemma 8.1

(Scan lemma) For any $R: B \leftarrow \mathrm{~F}(A, B)$ we have

$$
\text { term }^{+}([R \mathbb{D} \cdot([a c c \alpha \mathbb{D}) \supseteq \llbracket a c c R \mathbb{}] .
$$

## Proof

The proof is an exercise in fusion. We get the required result by showing

$$
\text { term }^{+}\left(\left[R D \cdot \text { acc } \alpha \quad \supseteq \operatorname { a c c } R \cdot F \left(\text {id }, \text { term }{ }^{+}([R D) .\right.\right.\right.
$$

In the argument that follows we make use of some laws of products that we have not formally stated:

$$
\left.\begin{array}{ll} 
& \text { term }^{+}([R]) \cdot \text { acc } \alpha \\
= & \{\text { definition of acc }\}
\end{array}\right\}
$$

### 8.5 Deforestation

The structure built up by an scan is usually not the final result of a computation; in practice, the labelled term that results is often evaluated by a catamorphism. This means that the labelled term need never be built up as a whole, for one can merge the process of its construction and its evaluation. This technique is very common in functional programming; it has been called deforestation by Wadler (1990), and Swierstra and De Moor (1992) speak of virtual data structures. Our final lemma shows how deforestation can be used in the present context:

## Lemma 8.2

(Deforestation) Let $R: B \leftarrow \mathrm{~F}^{+}(C, B)$ and $S: C \leftarrow \mathrm{~F}(A, C)$, so $\left[R \mathbb{D}: B \leftarrow\right.$ term $^{+} C$ and $\llbracket a c c S \rrbracket:$ term $^{+} C \leftarrow$ term $A$. Then

$$
([R] \cdot([\text { acc } S]) \supseteq \text { outl } \cdot(\text { Ireduce }(R, S)] \text {, }
$$

where reduce $(R, S):(B \times C) \leftarrow \mathrm{F}(A, B \times C)$ is given by

$$
\text { reduce }(R, S)=\langle R, \text { outr }\rangle \cdot\langle F(!, \text { outl }), S \cdot F(\text { id, outr })\rangle .
$$

Again the proof is an exercise in fusion, and we will not go into the details. If the final program is going to be evaluated in a lazy programming language, this lemma does not offer a real improvement in efficiency: the intermediate data structure in ([S]) • (acc R] never exists in its entirety anyway. It is probably for this reason that the above result was never stated in the theory of lists.

## 9 Segments and segment decomposition

Having dealt with all the necessary ingredients, we can now give the general version of the calculation given in section 3. First, we define segment $:$ term $A \leftarrow$ term $A$ by

$$
\text { segment }=\text { prune } \cdot \text { subterm }
$$

Now we have

## Theorem 9.1

(Segment Decomposition)
Let F be a pointed bifunctor with membership relations left and right. Suppose $f=\left[f_{0}, f_{1}\right]: B \leftarrow \mathrm{~F}(A, B)$ is monotonic under the preorder $R: B \leftarrow B$. Then $\max R \cdot \mathrm{P}(f]) \cdot \Lambda$ segment $\supseteq$ outl $\cdot([$ reduce $(S, T)]$,
where $S=\max R \cdot \Lambda($ left $\cup$ right $)$ and $T=\max R \cdot \Lambda\left[f_{0}, f_{0} \cup f_{1}\right]$.
Proof
We argue:

$$
\begin{aligned}
& \max R \cdot \mathrm{P}(f \mathrm{D} \cdot \Lambda \text { segment } \\
= & \{\text { definition of segment }\} \\
& \max R \cdot \mathrm{P}(f \mathrm{D} \cdot \Lambda(\text { prune } \cdot \text { subterm })
\end{aligned}
$$

$=\{(3)\}$
$\max R \cdot \mathrm{P}(f \mathrm{f} \mathrm{D} \cdot$ union $\cdot \mathrm{P} \Lambda$ prune $\cdot \Lambda$ subterm
$=\{(4)$ and functors $\}$
$\max R \cdot$ union $\cdot \mathrm{P}(\mathrm{P}(f f \mathrm{D} \cdot \Lambda$ prune $) \cdot \Lambda$ subterm
$\supseteq \quad\{(6)$ and functors $\}$
$\max R \cdot \mathrm{P}($ max $R \cdot \mathrm{P}(f f \mathrm{D} \cdot \Lambda$ prune $) \cdot \Lambda$ subterm
$\supseteq \quad\left\{\right.$ Horner's rule with $\left.X=\left[f_{0}, f_{0} \cup f_{1}\right]\right\}$
$\max R \cdot \mathrm{P}([\max R \cdot \Lambda X \mathrm{D} \cdot \Lambda$ subterm
$=\{(10)\}$
$\max R \cdot \mathrm{P}([\max R \cdot \Lambda X \mathrm{D} \cdot$ setify $\cdot$ subterms
$\supseteq \quad\{(7)\}$
$\max R \cdot$ setify $\cdot$ term $\left.^{+}{ }_{(\max R} \cdot \Lambda X\right] \cdot$ subterms
$\supseteq \quad\{$ Scan Lemma\} $\max R \cdot \operatorname{setify} \cdot(\mathbb{a c c}(\max R \cdot \Lambda X) \rrbracket$
$\supseteq \quad\{(8)\}$
$[(\max R \cdot \Lambda(l e f t \cup \operatorname{right}) \rrbracket \cdot(\mathbb{a c c}(\max R \cdot \Lambda X) D$
$\supseteq \quad\{$ Deforestation Lemma, definitions of $S$ and $T\}$
outl $\cdot($ reduce $(S, T))$ ).

Application of the segment decomposition theorem gives an efficient solution for the maximum segment sum problem on any type that allows the definition of sum.

## 10 Concluding remarks

We have demonstrated how much of the original theory of lists can be parameterised by an arbitrary data type. The result is, in our opinion, at least a linguistic improvement; the theory is no longer cluttered by the syntactic idiosyncracies of lists. It is debatable, however, whether by itself any mere linguistic improvement would justify the flood of definitions and results given above. What is of greater interest is the possibility that this style of generic programming can be applied to more challenging problems. An obvious candidate for further work is the so-called sliding tails lemma, which underlies all efficient pattern matching algorithms on lists. If this lemma can be parameterised by an arbitrary data type, the way is open for a generic theory of pattern matching. Such a generic theory is likely to benefit by the work of Backhouse (1992), who has shown how many theorems about regular algebra can be generalised to datatypes.
Finally, another important direction for future research is the design of a programming language in which data types are first-class citizens, in the sense that they can be passed as parameters to generic programs. It seems that research in the design
of functional programming languages is also heading in this direction; in particular the work of Jones (1993) on constructor classes is relevant in this connection.

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