## Chapter 3

## Recursion in the Pointfree Style

How useful from a programmer's point of view are the abstract concepts presented in the previous chapter? Recall that a table was presented - table 2.1 - which records an analogy between abstract type notation and the corresponding datastructures available in common, imperative languages.

This analogy will help in showing how to extend the abstract notation studied thus far towards a most important field of programming: recursion. This, however, will be preceeded by a simpler introduction to the subject rooted on very basic and intuitive notions of mathematics.

### 3.1 Motivation

Where do algorithms come from? From human imagination only? Surely not they actually emerge from mathematics. In a sense, in the same way one may say that hardware follows the laws of physics (eg. semiconductor electronics) one might say that software is governed by the laws of mathematics.

This section provides a naive introduction to algorithm analysis and synthesis by showing how a quite elementary class of algorithms - equivalent to for-loops in C or any other imperative language - arise from elementary properties of the underlying maths domain.

We start by showing how the arithmetic operation of multiplying two natural numbers (in $\mathbb{N}_{0}$ ) is a for-loop which emerges solely from the algebraic properties
of multiplication:

$$
\left\{\begin{array}{l}
a \times 0=0  \tag{3.1}\\
a \times 1=a \\
a \times(b+c)=a \times b+a \times c
\end{array}\right.
$$

These properties are known as the absorption, unit and distributive properties of multiplication, respectively.

Start by making $c:=1$ in the third (distributive) property, obtaining $a \times(b+$ 1) $=a \times b+a \times 1$, and then simplify. The second clause is useful in this simplification but it is not required in the final system of two equations,

$$
\left\{\begin{array}{l}
a \times 0=0  \tag{3.2}\\
a \times(b+1)=a \times b+a
\end{array}\right.
$$

since it is derivable from the remaining two, for $b:=0$ and property $0+a=a$ of addition.

System (3.2) is already a runnable program in a functional language such as Haskell (among others). The moral of this trivial exercise is that programs arise from the underlying maths, instead of being invented or coming out of the blue. Novices in functional programming do this kind of reasoning all the time without even noticing it, when writing their first programs. For instance, the function which computes discrete exponentials will scale up the same procedure, thanks to the properties

$$
\left\{\begin{array}{l}
a^{0}=1 \\
a^{1}=a \\
a^{b+c}=a^{b} \times a^{c}
\end{array}\right.
$$

where the program just developed for multiplication can be re-used, and so and so on.

Type-wise, the multiplication algorithm just derived for natural numbers is not immediate to generalize. Intuitively, it will diverge for $b$ a negative integer and for $b$ a real number less than 1 , at least. Argument $a$, however, does not seem to be constrained.

Indeed, the two arguments $a$ and $b$ will have different types in general. Let us see why and how. Starting by looking at infix operators $(\times)$ and $(+)$ as curried operators - recall section 2.15 - we can resort to the corresponding sections and write:

$$
\left\{\begin{array}{l}
(a \times) 0=0  \tag{3.3}\\
(a \times)(b+1)=(a+)((a \times) b)
\end{array}\right.
$$

It can be easily checked that

$$
\begin{equation*}
(a \times)=\text { for }(a+) 0 \tag{3.4}
\end{equation*}
$$

by introducing a for-loop combinator given by

$$
\left\{\begin{array}{l}
\text { for } f i 0=i  \tag{3.5}\\
\text { for } f i(n+1)=f(\text { for } f i n)
\end{array}\right.
$$

where $f$ is the loop-body and $i$ is the initialization value. In fact, (for $f i) n=f^{n} i$, that is, $f$ is iterated $n$ times over the initial value $i$.

For-loops are a primitive construct available in many programming languages. In C, for instance, one will write something like

```
int mul(int a, int n)
{
int s=0; int i;
for (i=1;i<n+1;i++) {s += a;}
return s;
};
```

for (the uncurried version of) for $(a+) 0$ loop.
To better understand this construct let us remove variables from both equations in (3.3) by lifting function application to function composition and lifting 0 to the "everywhere 0 " (constant) function:

$$
\left\{\begin{array}{l}
(a \times) \cdot \underline{0}=\underline{0} \\
(a \times) \cdot(+1)=(+a) \cdot(a \times)
\end{array}\right.
$$

Using the junc ("either") pointfree combinator we merge the two equations into a single one,

$$
[(a \times) \cdot \underline{0},(a \times) \cdot(+1)]=[\underline{0},(+a) \cdot(a \times)]
$$

— thanks to the Eq-+ rule (2.66) — then single out the common factor $(a \times)$ in the left hand side,

$$
(a \times) \cdot[\underline{0},(+1)]=[\underline{0},(+a) \cdot(a \times)]
$$

- thanks to +-fusion (2.42) - and finally do a similar fission operation on the other side,

$$
\begin{equation*}
(a \times) \cdot[\underline{0},(+1)]=[\underline{0},(+a)] \cdot(i d+(a \times)) \tag{3.6}
\end{equation*}
$$

— thanks to +-absorption (2.43).
As we already know, equalities of compositions are nicely drawn as diagrams. That of (3.6) is as follows:


Function $(+1)$ is the successor function succ on natural numbers. Type $A$ is any (non-empty) type. For the particular case of $A=1$, the diagram is more interesting, as $[\underline{0}$, succ $]$ becomes an isomorphism, telling a unique way of building natural numbers: ${ }^{1}$

## Every natural number in $\mathbb{N}_{0}$ either is 0 or the successor of another natural number.

We will denote such an isomorphism by $i n$ and its converse by out in the following version of the same diagram

capturing both the isomorphism and the $(a \times)$ recursive function. By solving the isomorphism equation out $\cdot i n=i d$ we easily obtain the definition of out, the converse of in ${ }^{2}$ :

$$
\begin{aligned}
& \text { out } 0=i_{1}() \\
& \operatorname{out}(n+1)=i_{2} n
\end{aligned}
$$

[^0]Finally, we generalize the target $\mathbb{N}_{0}$ to any non-empty type $B,(+a)$ to any function $B \xrightarrow{g} B$ and 0 to any constant $k$ in $B$ (this is why $B$ has to be non-empty). The corresponding generalization of $(a \times)$ is denoted by $f$ below:


It turns out that, given $k$ and $g$, there is a unique solution to the equation (in $f$ ) captured by the diagram: $f \cdot i n=[\underline{k}, g] \cdot(i d+f)$. We know this solution already, recall (3.5):

$$
f=\text { for } g k
$$

As we have seen earlier on, solution uniqueness is captured by universal properties. In this case we have the following property, which we will refer to by writing "for-loop-universal":

$$
\begin{equation*}
f=\text { for } g k \equiv f \cdot i n=[\underline{k}, g] \cdot(i d+f) \tag{3.7}
\end{equation*}
$$

From this property it is possible to infer a basic theory of for-loops. For instance, by making $f=i d$ and solving the for-loop-universal equation (3.7) for $g$ and $k$ we obtain the reflexion law:

$$
\begin{equation*}
\text { for succ } 0=i d \tag{3.8}
\end{equation*}
$$

This can be compared with the following (useless) program in C :

```
int id(int n)
{
    int s=0; int i;
    for (i=1;i<n+1;i++) {s += 1;}
    return s;
};
```

(Clearly, the value returned in $s$ is that of input $n$.)
More knowledge about for-loops can be extracted from (3.7). Later on we will show that these constructs are special cases of a more general concept termed
catamorphism. ${ }^{3}$ In the usual "banana-bracket" notation of catamorphisms, to be introduced later, the for-combinator will be written:

$$
\begin{equation*}
\text { for } g k=([\underline{k}, g] D \tag{3.9}
\end{equation*}
$$

In the sequel, we shall study the (more general) theory of catamorphisms and come back to for-loops as an instantiation. Then we will understand how more interesting for-loops can be synthesized, for instance those handling more than one "global variable", thanks to catamorphism theory (for instance, the mutual recursion laws).

As a generalization to what we've just seen happening between for-loops and natural numbers, it will be shown that a catamorphism is intimately connected to the data-structure it processes, for instance a finite list (sequence) or a binary tree. A good understanding of such structures is therefore required. We proceed to studying the list data structure first, wherefrom trees stem as natural extensions.

Exercise 3.1. Addition is known to be associative $(a+(b+c)=(a+b)+c)$ and have unit $0(a+0=a)$. Following the same strategy that was adopted above for $(a \times)$, show that

$$
\begin{equation*}
(a+)=\text { for succ } a \tag{3.10}
\end{equation*}
$$

Exercise 3.2. The following fusion-law

$$
\begin{equation*}
h \cdot(\text { for } g k)=\text { for } j(h k) \Leftarrow h \cdot g=j \cdot h \tag{3.11}
\end{equation*}
$$

can be derived from universal-property $(3.7)^{4}$. Since $(a+) \cdot i d=(a+)$, provide an alternative derivation of (3.10) using the fusion-law above.

Exercise 3.3. From (3.4) and fusion-law (3.11) infer: $(a *) \cdot$ succ $=$ for $a(a+)$.

[^1]Exercise 3.4. Show that $f=$ for $\underline{k} k$ and $g=$ for $i d k$ are the same program (function).

Exercise 3.5. Generic function $k=$ for $f i$ can be encoded in the syntax of $C$ by writing

```
int k(int n) {
    int r=i;
    int x;
    for (x=1;x<n+1;x++) {r=f(r);}
    return r;
};
```

for some predefined $f$. Encode the functions $f$ and $g$ of exercise 3.4 in $C$ and compare them.

### 3.2 From natural numbers to finite sequences

Let us consider a very common data-structure in programming: "linked-lists". In PASCAL one will write

```
L = ^N;
N = record
    first: A;
    next: ^N
end;
```

to specify such a data-structure $L$. This consists of a pointer to a node ( N ), where a node is a record structure which puts some predefined type $A$ together with a pointer to another node, and so on. In the C programming language, every $x \in L$ will be declared as $L x$ in the context of datatype definition

```
typedef struct N {
    A first;
    struct N *next;
} *L;
```

and so on.
What interests us in such "first year programming course" datatype declarations? Records and pointers have already been dealt with in table 2.1. So we can use this table to find the abstract version of datatype $L$, by replacing pointers by the " $1+\ldots$ " notation and records (structs) by the "... $\times \ldots$. . notation:

$$
\left\{\begin{align*}
L & =1+N  \tag{3.12}\\
N & =A \times(1+N)
\end{align*}\right.
$$

We obtain a system of two equations on unknowns $L$ and $N$, in which $L$ 's dependence on $N$ can be removed by substitution:

$$
\left.\left.\begin{array}{rl} 
& \left\{\begin{aligned}
L & =1+N \\
N & =A \times(1+N)
\end{aligned}\right. \\
\equiv & \begin{array}{rl}
\{\text { substituting } L \text { for } 1+N \text { in the second equation }\}
\end{array} \\
\equiv\left\{\begin{array}{rl}
L & =1+N \\
N & =
\end{array}+L \times L\right.
\end{array}\right\} \begin{array}{rl}
\{\text { substituting } A \times L \text { for } N \text { in the first equation }\}
\end{array}\right\}
$$

System (3.12) is thus equivalent to:

$$
\left\{\begin{align*}
L & =1+A \times L  \tag{3.13}\\
N & =A \times(1+N)
\end{align*}\right.
$$

Intuitively, $L$ abstracts the "possibly empty" linked-list of elements of type $A$, while $N$ abstracts the "non-empty" linked-list of elements of type $A$. Note that $L$ and $N$ are independent of each other, but also that each depends on itself. Can we solve these equations in a way such that we obtain "solutions" for $L$ and $N$, in the same way we do with school equations such as, for instance,

$$
\begin{equation*}
x=1+\frac{x}{2} \quad ? \tag{3.14}
\end{equation*}
$$

Concerning this equation, let us recall how we would go about it in school mathematics:

$$
\begin{aligned}
& x=1+\frac{x}{2} \\
& \equiv \quad\left\{\text { adding }-\frac{x}{2} \text { to both sides of the equation }\right\} \\
& x-\frac{x}{2}=1+\frac{x}{2}-\frac{x}{2} \\
& \equiv \quad\left\{-\frac{x}{2} \text { cancels } \frac{x}{2}\right\} \\
& x-\frac{x}{2}=1 \\
& \equiv \quad\{\text { multiplying both sides of the equation by } 2 \text { etc. }\} \\
& 2 \times x-x=2 \\
& \equiv \quad\{\text { subtraction }\} \\
& x=2
\end{aligned}
$$

We very quickly get solution $x=2$. However, many steps were omitted from the actual calculation. This unfolds into the longer sequence of more elementary steps which follows, in which notation $a-b$ abbreviates $a+(-b)$ and $\frac{a}{b}$ abbreviates $a \times \frac{1}{b}$, for $b \neq 0$ :

$$
\begin{aligned}
& x=1+\frac{x}{2} \\
& \equiv \quad\left\{\text { adding }-\frac{x}{2} \text { to both sides of the equation }\right\} \\
& x-\frac{x}{2}=\left(1+\frac{x}{2}\right)-\frac{x}{2} \\
& \equiv \quad\{+ \text { is associative }\} \\
& x-\frac{x}{2}=1+\left(\frac{x}{2}-\frac{x}{2}\right) \\
& \equiv \quad\left\{-\frac{x}{2} \text { is the additive inverse of } \frac{x}{2}\right\} \\
& x-\frac{x}{2}=1+0 \\
& \equiv \quad\{0 \text { is the unit of addition }\} \\
& x-\frac{x}{2}=1
\end{aligned}
$$

$$
\begin{aligned}
& \equiv \quad\{\text { multiplying both sides of the equation by } 2\} \\
& 2 \times\left(x-\frac{x}{2}\right)=2 \times 1 \\
& \equiv \quad\{1 \text { is the unit of multiplication }\} \\
& 2 \times\left(x-\frac{x}{2}\right)=2 \\
& \equiv \quad\{\text { multiplication distributes over addition }\} \\
& 2 \times x-2 \times \frac{x}{2}=2 \\
& \equiv \quad\left\{2 \text { cancels its inverse } \frac{1}{2}\right\} \\
& 2 \times x-1 \times x=2 \\
& \equiv \quad\{\text { multiplication distributes over addition }\} \\
& (2-1) \times x=2 \\
& \equiv \quad\{2-1=1 \text { and } 1 \text { is the unit of multiplication }\} \\
& x=2
\end{aligned}
$$

Back to (3.13), we would like to submit each of the equations, e.g.

$$
\begin{equation*}
L=1+A \times L \tag{3.15}
\end{equation*}
$$

to a similar reasoning. Can we do it? The analogy which can be found between this equation and (3.14) goes beyond pattern similarity. From chapter 2 we know that many properties required in the reasoning above hold in the context of (3.15), provided the " $=$ " sign is replaced by the " $\cong$ " sign, that of set-theoretical isomorphism. Recall that, for instance, + is associative (2.48), 0 is the unit of addition (2.55), 1 is the unit of multiplication (2.57), multiplication distributes over addition (2.52) etc. Moreover, the first step above assumed that addition is compatible (monotonic) with respect to equality,

$$
\begin{array}{ccc}
a & = & b \\
c & = & d \\
\hline a+c & = & b+d
\end{array}
$$

a fact which still holds when numeric equality gives place to isomorphism and numeric addition gives place to coproduct:

$$
\begin{array}{ccc}
A & \cong B \\
C & \cong D \\
\hline A+C & \cong B+D
\end{array}
$$

— recall (2.46) for isos $f$ and $g$.
Unfortunately, the main steps in the reasoning above are concerned with two basic cancellation properties

$$
\begin{aligned}
x+b=c & \equiv x=c-b \\
x \times b=c & \equiv x=\frac{c}{b} \quad(b \neq 0)
\end{aligned}
$$

which hold about numbers but do not hold about datatypes. In fact, neither products nor coproducts have arbitrary inverses ${ }^{5}$, and so we cannot "calculate by cancellation". How do we circumvent this limitation?

Just think of how we would have gone about (3.14) in case we didn't know about the cancellation properties: we would be bound to the $x$ by $1+\frac{x}{2}$ substitution plus the other properties. By performing such a substitution over and over again we would obtain...

$$
\begin{aligned}
& x=1+\frac{x}{2} \\
& \equiv \quad\left\{x \text { by } 1+\frac{x}{2} \text { substitution followed by simplification }\right\} \\
& x=1+\frac{1+\frac{x}{2}}{2}=1+\frac{1}{2}+\frac{x}{4} \\
& \equiv \quad\{\text { the same as above }\} \\
& x=1+\frac{1}{2}+\frac{1+\frac{x}{2}}{4}=1+\frac{1}{2}+\frac{1}{4}+\frac{x}{8} \\
& \equiv \quad\{\text { over and over again, } n \text {-times }\} \\
& \equiv \quad\{\text { simplification }\} \\
& x=\sum_{i=0}^{n} \frac{1}{2^{i}}+\frac{x}{2^{n+1}} \\
& \equiv \quad\{\text { sum of } n \text { first terms of a geometric progression }\} \\
& x=\left(2-\frac{1}{2^{n}}\right)+\frac{x}{2^{n+1}}
\end{aligned}
$$

[^2]\[

$$
\begin{gathered}
\equiv \quad \begin{array}{c}
\{\text { let } n \rightarrow \infty\} \\
\\
x= \\
\equiv \quad(2-0)+0 \\
\quad\{\text { simplification }\} \\
x=2
\end{array}
\end{gathered}
$$
\]

Clearly, this is a much more complicated way of finding solution $x=2$ for equation (3.14). But we would have loved it in case it were the only known way, and this is precisely what happens with respect to (3.15). In this case we have:

$$
\begin{aligned}
& L=1+A \times L \\
& \equiv \quad\{\text { substitution of } 1+A \times L \text { for } L\} \\
& L=1+A \times(1+A \times L) \\
& \equiv \quad\{\text { distributive property (2.52) }\} \\
& L \cong 1+A \times 1+A \times(A \times L) \\
& \equiv \quad\{\text { unit of product (2.57) and associativity of product (2.34) }\} \\
& L \cong 1+A+(A \times A) \times L \\
& \equiv \quad\{\text { by }(2.96),(2.98) \text { and (2.102) }\} \\
& L \cong A^{0}+A^{1}+A^{2} \times L \\
& \equiv \quad\{\text { another substitution as above and similar simplifications }\} \\
& L \cong A^{0}+A^{1}+A^{2}+A^{3} \times L \\
& \equiv \quad\{\operatorname{after}(n+1) \text {-many similar steps }\} \\
& L \cong \sum_{i=0}^{n} A^{i}+A^{n+1} \times L
\end{aligned}
$$

Bearing a large $n$ in mind, let us deliberately (but temporarily) ignore term $A^{n+1} \times L$. Then $L$ will be isomorphic to the sum of $n$-many contributions $A^{i}$,

$$
L \cong \sum_{i=0}^{n} A^{i}
$$

each of them consisting of $i$-long tuples, or sequences, of values of $A$. (Number $i$ is said to be the length of any sequence in $A^{i}$.) Such sequences will be denoted by enumerating their elements between square brackets, for instance the empty
sequence [] which is the only inhabitant in $A^{0}$, the two element sequence [ $a_{1}, a_{2}$ ] which belongs to $A^{2}$ provided $a_{1}, a_{2} \in A$, and so on. Note that all such contributions are mutually disjoint, that is, $A^{i} \cap A^{j}=\emptyset$ wherever $i \neq j$. (In other words, a sequence of length $i$ is never a sequence of length $j$, for $i \neq j$.) If we join all contributions $A^{i}$ into a single set, we obtain the set of all finite sequences on $A$, denoted by $A^{\star}$ and defined as follows:

$$
\begin{equation*}
A^{\star} \stackrel{\text { def }}{=} \bigcup_{i \geqslant 0} A^{i} \tag{3.16}
\end{equation*}
$$

The intuition behind taking the limit in the numeric calculation above was that term $\frac{x}{2^{n+1}}$ was getting smaller and smaller as $n$ went larger and larger and, "in the limit", it could be ignored. By analogy, taking a similar limit in the calculation just sketched above will mean that, for a "sufficiently large" $n$, the sequences in $A^{n}$ are so long that it is very unlikely that we will ever use them! So, for $n \rightarrow \infty$ we obtain

$$
L \cong \sum_{i=0}^{\infty} A^{i}
$$

Because $\sum_{i=0}^{\infty} A^{i}$ is isomorphic to $\bigcup_{i=0}^{\infty} A^{i}$ (see exercise 2.31), we finally have:

$$
L \cong A^{\star}
$$

All in all, we have obtained $A^{\star}$ as a solution to equation (3.15). In other words, datatype $L$ is isomorphic to the datatype which contains all finite sequences of some predefined datatype $A$. This corresponds to the Haskell [a] datatype, in general. Recall that we started from the "linked-list datatype" expressed in PASCAL or C. In fact, wherever the C programmer thinks of linked-lists, the HASKELL programmer will think of finite sequences.

But, what does equation (3.15) mean in fact? Is $A^{\star}$ the only solution to this equation? Back to the numeric field, we know of equations which have more than one solution - for instance $x=\frac{x^{2}+3}{4}$, which admits two solutions 1 and 3 -, which have no solution at all - for instance $x=x+1$ - , or which admit an infinite number of - for instance $x=x$.

We will address these topics in the next section about inductive datatypes and in chapter 9 , where the formal semantics of recursion will be made explicit. This is where the "limit" constructions used informally in this section will be shown to make sense.

### 3.3 Introducing inductive datatypes

Datatype $L$ as defined by (3.15) is said to be recursive because $L$ "recurs" in the definition of $L$ itself ${ }^{6}$. From the discussion above, it is clear that set-theoretical equality " $=$ " in this equation should give place to set-theoretical isomorphism ("ㅡ"):

$$
\begin{equation*}
L \cong 1+A \times L \tag{3.17}
\end{equation*}
$$

Which isomorphism $L \longleftarrow_{\longleftarrow}^{i n} 1+A \times L$ do we expect to witness (3.15)? This will depend on which particular solution to (3.15) we are thinking of. So far we have seen only one, $A^{\star}$. By recalling the notion of algebra of a datatype (section 2.18), so we may rephrase the question as: which algebra

$$
A^{\star} \stackrel{i n}{\longleftarrow} 1+A \times A^{\star}
$$

do we expect to witness the tautology which arises from (3.15) by replacing unknown $L$ with solution $A^{\star}$, that is

$$
A^{\star} \cong 1+A \times A^{\star} \quad ?
$$

It will have to be of the form $\mathrm{in}=\left[i n_{1}, i n_{2}\right]$ as depicted by the following diagram:


Arrows $i n_{1}$ and $i n_{2}$ can be guessed rather intuitively: $i n_{1}=[]$, which will express the "NIL pointer" by the empty sequence, at $A^{\star}$ level, and $i n_{2}=\mathrm{cons}$, where cons is the standard "left append" sequence constructor, which we for the moment introduce rather informally as follows:

$$
\begin{align*}
& \operatorname{cons}: A \times A^{\star} \rightarrow A^{\star} \\
& \operatorname{cons}\left(a,\left[a_{1}, \ldots, a_{n}\right]\right)=\left[a, a_{1}, \ldots, a_{n}\right] \tag{3.19}
\end{align*}
$$

In a diagram:


[^3]Of course, for in to be iso it needs to have an inverse, which is not hard to guess,

$$
\begin{equation*}
\text { out } \stackrel{\text { def }}{=}(!+\langle h d, t l\rangle) \cdot\left(=\left[=_{[]} ?\right)\right. \tag{3.21}
\end{equation*}
$$

where sequence operators $h d$ (head of a nonempty sequence) and $t l$ (tail of a nonempty sequence) are (again informally) described as follows:

$$
\begin{align*}
& h d: A^{\star} \rightarrow A  \tag{3.22}\\
& h d\left[a_{1}, a_{2}, \ldots, a_{n}\right]=a_{1} \\
& t l: A^{\star} \rightarrow A^{\star}  \tag{3.23}\\
& t l\left[a_{1}, a_{2}, \ldots, a_{n}\right]=\left[a_{2}, \ldots, a_{n}\right]
\end{align*}
$$

Showing that in and out are each other inverses is not a hard task either:

$$
\begin{aligned}
& \text { in } \cdot \text { out }=i d \\
& \equiv \quad\{\text { definitions of } \text { in } \text { and out }\} \\
& \text { [[] }, \text { cons }] \cdot(!+\langle h d, t l\rangle) \cdot\left(=_{[]} ?\right)=i d \\
& \equiv \quad\{+ \text {-absorption (2.43) and (2.15) }\} \\
& {[\underline{[]}, \text { cons } \cdot\langle h d, t l\rangle] \cdot\left(=_{[]} ?\right)=i d} \\
& \equiv \quad\{\text { property of sequences: } \operatorname{cons}(h d s, t l s)=s\} \\
& {[\underline{]}, i d] \cdot(=[] ?)=i d} \\
& \equiv \quad\{\text { going pointwise (2.69) }\} \\
& \left\{\begin{array}{rl}
=[] a & \Rightarrow[[], i d]\left(i_{1} a\right) \\
\neg(=[] a) & \Rightarrow[\underline{[]]}, i d]\left(i_{2} a\right)
\end{array}=a\right. \\
& \equiv\{+ \text {-cancellation (2.40) }\} \\
& \left\{\begin{aligned}
=_{[]} a & \Rightarrow \frac{[]}{} a \\
\neg\left(=_{[]} a\right) & \Rightarrow \frac{\text { id } a}{}=a
\end{aligned}\right. \\
& \equiv \quad\{a=[] \text { in one case and identity function (2.9) in the other }\} \\
& \left\{\begin{array}{ccc}
a=[] & \Rightarrow a \\
\neg(a=[]) & \Rightarrow & a
\end{array}=a\right. \\
& \equiv \quad\{\text { property }(p \rightarrow f, f)=f \text { holds }\}
\end{aligned}
$$

A comment on the particular choice of terminology above: symbol in suggests that we are going inside, or constructing (synthesizing) values of $A^{\star}$; symbol out suggests that we are going out, or destructing (analyzing) values of $A^{\star}$. We shall often resort to this duality in the sequel.

Are there more solutions to equation (3.17)? In trying to implement this equation, a HASKELL programmer could have written, after the declaration of type $A$, the following datatype declaration:

```
data L = Nil () | Cons (A,L)
```

which, as we have seen in section 2.18 , can be written simply as

$$
\begin{equation*}
\text { data } L=N i l \mid \operatorname{Cons}(A, L) \tag{3.24}
\end{equation*}
$$

and generates diagram

leading to algebra $i n^{\prime}=[\underline{N i l}$, Cons $]$.
HASKELL seems to have generated another solution for the equation, which it calls $L$. To avoid the inevitable confusion between this symbol denoting the newly created datatype and symbol $L$ in equation (3.17), which denotes a mathematical variable, let us use symbol T to denote the former ( T stands for "type"). This can be coped with very simply by writing $T$ instead of $L$ in (3.24):

$$
\begin{equation*}
\text { data } T=N i l \mid \operatorname{Cons}(A, T) \tag{3.26}
\end{equation*}
$$

In order to make $T$ more explicit, we will write $i n_{\mathrm{T}}$ instead of $i n^{\prime}$.
Some questions are on demand at this point. First of all, what is datatype T?
What are its inhabitants? Next, is $\mathrm{T} \stackrel{i n_{\mathrm{T}}}{\leftarrow} 1+A \times \mathrm{T}$ an iso or not?
HASKELL will help us to answer these questions. Suppose that $A$ is a primitive numeric datatype, and that we add deriving Show to (3.26) so that we can "see" the inhabitants of the $T$ datatype. The information associated to $T$ is thus:

```
Main> :i T
-- type constructor
data T
```

```
-- constructors:
Nil :: T
Cons :: (A,T) -> T
-- instances:
instance Show T
instance Eval T
```

By typing Nil

```
Main> Nil
Nil :: T
```

we confirm that $N i l$ is itself an inhabitant of T, and by typing Cons

```
Main> Cons
<<function>> :: (A,T) -> T
```

we realize that Cons is not so (as expected), but it can be used to build such inhabitants, for instance:

```
Main> Cons(1,Nil)
Cons (1,Nil) :: T
```

or

```
Main> Cons(2,Cons(1,Nil))
Cons (2,Cons (1,Nil)) :: T
```

etc. We conclude that expressions involving Nil and Cons are inhabitants of type T. Are these the only ones? The answer is yes because, by design of the Haskell language, the constructors of type T will remain fixed once its declaration is interpreted, that is, no further constructor can be added to T . Does $i n_{\mathrm{T}}$ have an inverse? Yes, its inverse is coalgebra

$$
\begin{align*}
& \text { out }_{\mathrm{T}}: \mathrm{T} \rightarrow 1+A \times \mathrm{T} \\
& \text { out }_{\mathrm{T}} \text { Nil }=i_{1} \mathrm{NIL}  \tag{3.27}\\
& \text { out }_{\mathrm{T}}(\operatorname{Cons}(a, l))=i_{2}(a, l)
\end{align*}
$$

which can be straightforwardly encoded in HASKELL using the Either realization of + (recall sections 2.9 and 2.18):

```
outT :: T -> Either () (A,T)
outT Nil = Left ()
outT (Cons(a,l)) = Right(a,l)
```

In summary, isomorphism

holds, where datatype T is inhabited by symbolic expressions which we may visualize very conveniently as trees, for instance

picturing expression $\operatorname{Cons}(2, \operatorname{Cons}(1, \mathrm{Nil}))$. Nil is the empty tree and Cons may be regarded as the operation which adds a new root and a new branch, say $a$, to a tree $t$ :


The choice of symbols T , Nil and Cons was rather arbitrary in (3.26). Therefore, an alternative declaration such as, for instance,

$$
\begin{equation*}
\text { data } U=\text { Stop } \mid \operatorname{Join}(A, U) \tag{3.29}
\end{equation*}
$$

would have been perfectly acceptable, generating another solution for the equation under algebra $[$ Stop, Join $]$. It is easy to check that (3.29) is but a renaming of Nil to Stop and of Cons to Join. Therefore, both datatypes are isomorphic, or "abstractly the same".

Indeed, any other datatype $X$ inductively defined by a constant and a binary constructor accepting $A$ and $X$ as parameters will be a solution to the equation. Because we are just renaming symbols in a consistent way, all such solutions are abstractly the same. All of them capture the abstract notion of a list of symbols.

We wrote "inductively" above because the set of all expressions (trees) which inhabit the type is defined by induction. Such types are called inductive and we shall have a lot more to say about them in chapter 9 .

Exercise 3.6. Obviously,

```
either (const []) (:)
```

does not work as a HASKELL realization of the mediating arrow in diagram (3.20). What do you need to write instead?

### 3.4 Observing an inductive datatype

Suppose that one is asked to express a particular observation of an inductive such as T (3.26), that is, a function of signature $B \stackrel{f}{\leftrightarrows} \mathrm{~T}$ for some target type $B$. Suppose, for instance, that $A$ is $\mathbb{N}_{0}$ (the set of all non-negative integers) and that we want to add all elements which occur in a T-list. Of course, we have to ensure that addition is available in $\mathbb{N}_{0}$,

$$
\begin{aligned}
& a d d: \mathbb{N}_{0} \times \mathbb{N}_{0} \rightarrow \mathbb{N}_{0} \\
& \operatorname{add}(x, y) \stackrel{\text { def }}{=} x+y
\end{aligned}
$$

and that $0 \in \mathbb{N}_{0}$ is a value denoting "the addition of nothing". So constant arrow $\mathbb{N}_{0}<\underline{0} 1$ is available. Of course, $\operatorname{add}(0, x)=a d d(x, 0)=x$ holds, for all $x \in \mathbb{N}_{0}$. This property means that $\mathbb{N}_{0}$, together with operator add and constant 0 , forms a monoid, a very important algebraic structure in computing which will be exploited intensively later in this book. The following arrow "packaging" $\mathbb{N}_{0}$, add and $\underline{0}$,

$$
\begin{equation*}
\mathbb{N}_{0} \leftarrow \stackrel{[\underline{0}, a d d]}{\longleftarrow} 1+\mathbb{N}_{0} \times \mathbb{N}_{0} \tag{3.30}
\end{equation*}
$$

is a convenient way to express such a structure. Combining this arrow with the algebra

$$
\begin{equation*}
\mathrm{T} \stackrel{i n_{\mathrm{T}}}{\leftrightarrows} 1+\mathbb{N}_{0} \times \mathrm{T} \tag{3.31}
\end{equation*}
$$

which defines T , and the function $f$ we want to define, the target of which is $B=\mathbb{N}_{0}$, we get the almost closed diagram which follows, in which only the dashed arrow is yet to be filled in:


We know that $i n_{\mathrm{T}}=[\underline{N i l}$, Cons $]$. A pattern for the missing arrow is not difficult to guess: in the same way $f$ bridges T and $\mathbb{N}_{0}$ on the lefthand side, it will do the same job on the righthand side. So pattern $\cdots+\cdots \times f$ comes to mind (recall section 2.10 ), where the ". . ." are very naturally filled in by identity functions. All in all, we obtain diagram

which pictures the following property of $f$

$$
\begin{equation*}
f \cdot[\underline{N i l}, C o n s]=[\underline{0}, a d d] \cdot(i d+i d \times f) \tag{3.34}
\end{equation*}
$$

and is easy to convert to pointwise notation:

$$
\left.\begin{array}{rl} 
& f \cdot[\underline{N i l}, C o n s]=[\underline{0}, a d d] \cdot(i d+i d \times f) \\
\equiv & \quad\{(2.42) \text { on the lefthand side, }(2.43) \text { and identity } i d \text { on the righthand side }\} \\
\equiv & {[f \cdot \underline{N i l}, f \cdot C o n s]=[\underline{0}, a d d \cdot(i d \times f)]} \\
& \{\text { either } \text { structural equality }(2.66)\}
\end{array}\right\} \begin{aligned}
& f \cdot \underline{N i l}=\underline{0} \\
& f \cdot \overline{C o n s}=a d d \cdot(i d \times f)
\end{aligned}
$$

$$
\begin{aligned}
& \equiv \quad\{\text { going pointwise }\} \\
& \left\{\begin{array}{l}
(f \cdot \underline{N i l}) x=\underline{0} x \\
(f \cdot \overline{\text { Cons }})(a, x)=(a d d \cdot(i d \times f))(a, x)
\end{array}\right. \\
& \equiv \quad\{\text { composition (2.6), constant (2.12), product (2.24) and definition of add }\} \\
& \left\{\begin{array}{l}
f \text { Nil }=0 \\
f(\operatorname{Cons}(a, x))=a+f x
\end{array}\right.
\end{aligned}
$$

Note that we could have used out $\mathrm{T}_{\mathrm{T}}$ in diagram (3.32),

obtaining another version of the definition of $f$,

$$
\begin{equation*}
f=[\underline{0}, a d d] \cdot(i d+i d \times f) \cdot \text { out }_{\mathrm{T}} \tag{3.36}
\end{equation*}
$$

which would lead to exactly the same pointwise recursive definition:

$$
\begin{aligned}
& f=[\underline{0}, a d d] \cdot(i d+i d \times f) \cdot \text { out }_{\boldsymbol{T}} \\
& \equiv \quad\{(2.43) \text { and identity } i d \text { on the righthand side }\} \\
& f=[\underline{0}, a d d \cdot(i d \times f)] \cdot \text { out }_{\top} \\
& \equiv \quad\left\{\text { going pointwise on out }{ }_{\top}(3.27)\right\} \\
& \left\{\begin{array}{l}
f \text { Nil }=\left([\underline{0}, \text { add } \cdot(i d \times f)] \cdot \text { out }_{\mathrm{T}}\right) \text { Nil }
\end{array}\right. \\
& \left\{f(\operatorname{Cons}(a, x))=\left([\underline{0}, a d d \cdot(i d \times f)] \cdot \text { out }_{\mathrm{T}}\right)(a, x)\right. \\
& \equiv \quad\left\{\text { definition of out }{ }_{\top}(3.27)\right\} \\
& \left\{\begin{array}{l}
f N i l=\left([\underline{0}, a d d \cdot(i d \times f)] \cdot i_{1}\right) N i l \\
f(\operatorname{Cons}(a, x))=([0, a d d \cdot(i d \times f)]
\end{array}\right. \\
& \equiv \quad\{+ \text {-cancellation (2.40) }\} \\
& \left\{\begin{array}{l}
f \text { Nil }=\underline{0} \text { Nil } \\
f(\operatorname{Cons}(a, x))=(a d d \cdot(i d \times f))(a, x)
\end{array}\right. \\
& \equiv \quad\{\text { simplification }\} \\
& \left\{\begin{array}{l}
f \text { Nil }=0 \\
f(\operatorname{Cons}(a, x))=a+f x
\end{array}\right.
\end{aligned}
$$

Pointwise $f$ mirrors the structure of type T in having has many definition clauses as constructors in $T$. Such functions are said to be defined by induction on the structure of their input type. If we repeat this calculation for $\mathbb{N}_{0}{ }^{\star}$ instead of $T$, that is, for

$$
\text { out }=(!+\langle h d, t l\rangle) \cdot\left(=_{[]} ?\right)
$$

— recall (3.21) — taking place of $o u t_{\mathrm{T}}$, we get a "more algorithmic" version of $f$ :

$$
\begin{aligned}
& f=[\underline{0}, a d d] \cdot(i d+i d \times f) \cdot(!+\langle h d, t l\rangle) \cdot\left(=_{[]} ?\right) \\
& \equiv \quad\{+ \text {-functor (2.44), identity and } \times \text {-absorption (2.27) }\} \\
& f=[\underline{0}, a d d] \cdot(!+\langle h d, f \cdot t l\rangle) \cdot\left(=_{[]} ?\right) \\
& \equiv \quad\{+ \text {-absorption (2.43) and constant } \underline{0}\} \\
& f=[\underline{0}, a d d \cdot\langle h d, f \cdot t l\rangle] \cdot\left(=_{[]} ?\right) \\
& \equiv \quad\left\{\text { going pointwise on guard }=_{[]}\right. \text {? (2.69) and simplifying \}} \\
& f l=\left\{\begin{array}{cl}
l=[] & \Rightarrow \underline{0} l \\
\neg(l=[]) & \Rightarrow(a d d \cdot\langle h d, f \cdot t l\rangle) l
\end{array}\right. \\
& \equiv \quad\{\text { simplification }\} \\
& f l=\left\{\begin{array}{cl}
l=[] & \Rightarrow 0 \\
\neg(l=[]) & \Rightarrow h d l+f(t l l)
\end{array}\right.
\end{aligned}
$$

The outcome of this calculation can be encoded in HASKELL syntax as

$$
\begin{aligned}
& f l \mid l \equiv[]=0 \\
& \quad \mid \text { otherwise }=\text { head } l+f(\text { tail } l)
\end{aligned}
$$

or

$$
f l=\text { if } l \equiv[] \text { then } 0 \text { else head } l+f(\text { tail } l)
$$

both requiring the equality predicate $\equiv$ and destructors head and tail.

### 3.5 Synthesizing an inductive datatype

The issue which concerns us in this section dualizes what we have just dealt with: instead of analyzing or observing an inductive type such as T (3.26), we want to be
able to synthesize (generate) particular inhabitants of T. In other words, we want to be able to specify functions with signature $B \xrightarrow{f} \mathrm{~T}$ for some given source type $B$. Let $B=\mathbb{N}_{0}$ and suppose we want $f$ to generate, for a given natural number $n>0$, the list containing all numbers less or equal to $n$ in decreasing order

$$
\operatorname{Cons}(n, \operatorname{Cons}(n-1, \operatorname{Cons}(\ldots, N i l)))
$$

or the empty list Nil, in case $n=0$.
Let us try and draw a diagram similar to (3.35) applicable to the new situation. In trying to "re-use" this diagram, it is immediate that arrow $f$ should be reversed. Bearing duality in mind, we may feel tempted to reverse all arrows just to see what happens. Identity functions are their own inverses, and $i n_{\mathrm{T}}$ takes the place of out ${ }_{\mathrm{T}}$ :


Interestingly enough, the bottom arrow is the one which is not obvious to reverse, meaning that we have to "invent" a particular destructor of $\mathbb{N}_{0}$, say

$$
\mathbb{N}_{0} \xrightarrow{g} 1+\mathbb{N}_{0} \times \mathbb{N}_{0}
$$

fitting in the diagram and generating the particular computational effect we have in mind. Once we do this, a recursive definition for $f$ will pop out immediately,

$$
\begin{equation*}
f=i n_{\mathrm{T}} \cdot(i d+i d \times f) \cdot g \tag{3.37}
\end{equation*}
$$

which is equivalent to:

$$
\begin{equation*}
f=[\underline{N i l}, \text { Cons } \cdot(i d \times f)] \cdot g \tag{3.38}
\end{equation*}
$$

Because we want $f 0=$ Nil to hold, $g$ (the actual generator of the computation) should distinguish input 0 from all the others. One thus decomposes $g$ as follows,

$$
\mathbb{N}_{0} \xlongequal[g]{\stackrel{=0}{\longrightarrow} \mathbb{N}_{0}+\mathbb{N}_{0} \xrightarrow{!+h} 1}+\mathbb{N}_{0} \times \mathbb{N}_{0}
$$

leaving $h$ to fill in. This will be a split providing, on the lefthand side, for the value to be Cons'ed to the output and, on the righthand side, for the "seed" to the
next recursive call. Since we want the output values to be produced contiguously and in decreasing order, we may define $h=\langle i d$, pred $\rangle$ where, for $n>0$,

$$
\begin{equation*}
\operatorname{pred} n \stackrel{\text { def }}{=} n-1 \tag{3.39}
\end{equation*}
$$

computes the predecessor of $n$. Altogether, we have synthesized

$$
\begin{equation*}
g=(!+\langle i d, \text { pred }\rangle) \cdot\left(==_{0} ?\right) \tag{3.40}
\end{equation*}
$$

Filling this in (3.38) we get

$$
\begin{aligned}
& f=[\underline{\text { Nil }}, \text { Cons } \cdot(i d \times f)] \cdot(!+\langle i d, \text { pred }\rangle) \cdot\left(==_{0} ?\right) \\
& \equiv \quad\{+ \text {-absorption (2.43) followed by } \times \text {-absorption (2.27) etc. }\} \\
& f=[\underline{\text { Nil }}, \text { Cons } \cdot\langle i d, f \cdot \text { pred }\rangle] \cdot\left(==_{0} ?\right) \\
& \equiv \quad\left\{\text { going pointwise on guard }=_{0} ?(2.69)\right. \text { and simplifying \}} \\
& f n=\left\{\begin{array}{ccl}
n=0 & \Rightarrow \text { Nil } \\
\neg(n=0) & \Rightarrow \quad \operatorname{Cons}(n, f(n-1))
\end{array}\right.
\end{aligned}
$$

which matches the function we had in mind:

$$
\begin{aligned}
& f n \\
& \quad \mid n \equiv 0=\text { Nil } \\
& \quad \mid \text { otherwise }=\text { Cons }(n, f(n-1))
\end{aligned}
$$

We shall see briefly that the constructions of the $f$ function adding up a list of numbers in the previous section and, in this section, of the $f$ function generating a list of numbers are very standard in algorithm design and can be broadly generalized. Let us first introduce some standard terminology.

### 3.6 Introducing (list) catas, anas and hylos

Suppose that, back to section 3.4, we want to multiply, rather than add, the elements occurring in lists of type T (3.26). How much of the program synthesis effort presented there can be reused in the design of the new function?

It is intuitive that only the bottom arrow $\mathbb{N}_{0} \stackrel{[0, \text { add }]}{\longleftarrow} 1+\mathbb{N}_{0} \times \mathbb{N}_{0}$ of diagram (3.35) needs to be replaced, because this is the only place where we can
specify that target datatype $\mathbb{N}_{0}$ is now regarded as the carrier of another (multiplicative rather than additive) monoidal structure,

$$
\begin{equation*}
\mathbb{N}_{0} \longleftarrow{ }^{[1, m u l]} 1+\mathbb{N}_{0} \times \mathbb{N}_{0} \tag{3.41}
\end{equation*}
$$

for $\operatorname{mul}(x, y) \stackrel{\text { def }}{=} x y$. We are saying that the argument list is now to be reduced by the multiplication operator and that output value 1 is expected as the result of "nothing left to multiply".

Moreover, in the previous section we might have wanted our number-list generator to produce the list of even numbers smaller than a given number, in decreasing order (see exercise 3.9). Intuition will once again help us in deciding that only arrow $g$ in (3.37) needs to be updated.

The following diagrams generalize both constructions by leaving such bottom arrows unspecified,

and express their duality ( $c f$. the directions of the arrows). It so happens that, for each of these diagrams, $f$ is uniquely dependent on the $g$ arrow, that is to say, each particular instantiation of $g$ will determine the corresponding $f$. So both $g$ s can be regarded as "seeds" or "genetic material" of the $f$ functions they uniquely define 7.

Following the standard terminology, we express these facts by writing $f=$ $(g)$ with respect to the lefthand side diagram and by writing $f=[(g)]$ with respect to the righthand side diagram. Read $(g)$ as "the T-catamorphism induced by $g$ " and $[(g)]$ as "the T-anamorphism induced by $g$ ". This terminology is derived from the Greek words $\kappa \alpha \tau \alpha$ (cata) and $\alpha \nu \alpha$ (ana) meaning, respectively, "downwards" and "upwards" (compare with the direction of the $f$ arrow in each diagram). The exchange of parentheses "( )" and "[ ]" in double parentheses "( )" and "[ )]" is aimed at expressing the duality of both concepts.

We shall have a lot to say about catamorphisms and anamorphisms of a given type such as T . For the moment, it suffices to say that

[^4]- the T-catamorphism induced by $B \stackrel{g}{\Vdash^{g}} 1+\mathbb{N}_{0} \times B$ is the unique func-


$$
\begin{equation*}
(g \mid)=g \cdot(i d+i d \times(g g \mid)) \cdot \text { out }_{\mathrm{T}} \tag{3.43}
\end{equation*}
$$

which is the same as

$$
\begin{equation*}
(g \mid) \cdot i n_{\mathrm{T}}=g \cdot(i d+i d \times(g \mid)) \tag{3.44}
\end{equation*}
$$

- given $B \xrightarrow{g} 1+\mathbb{N}_{0} \times B$ the T -anamorphism induced by $g$ is the unique function $B \xrightarrow{[(g)]} \mathrm{T}$ which obeys to property (or is defined by)

$$
\begin{equation*}
[(g)]=i n_{\mathbf{T}} \cdot(i d+i d \times[(g)]) \cdot g \tag{3.45}
\end{equation*}
$$

From (3.42) it can be observed that T can act as a mediator between any T -
 $\mathrm{T} \stackrel{L}{[(h)]} C$, for some $C \xrightarrow{h} 1+\mathbb{N}_{0} \times C$. In other words, a T-catamorphism call always observe (consume) the output of a T -anamorphism. The latter produces a list of $\mathbb{N}_{0} \mathrm{~s}$ which is consumed by the former. This is depicted in the diagram which follows:


What can we say about the $(g) \cdot[(h)]$ composition? It is a function from $C$ to $B$ which resorts to T as an intermediate data-structure and can be subject to the following calculation (cf. outermost rectangle in (3.46)):

$$
\left.\left.\begin{array}{rl} 
& (g \mid) \cdot[(h)]=g \cdot(i d+i d \times(g \mid)) \cdot(i d+i d \times[(h)]) \cdot h \\
\equiv & \{+ \text {-functor }(2.44)\}
\end{array}\right\} \quad(g \mid) \cdot[(h)]=g \cdot((i d \cdot i d)+(i d \times(g)) \cdot(i d \times[(h)))) \cdot h\right\}
$$

This calculation shows how to define $C \lessdot \frac{(g) \cdot[(h)]}{} B$ in one go, that is to say, doing without any intermediate data-structure:


As an example, let us see what comes out of $(g \mid) \cdot[(h)]$ for $h$ and $g$ respectively given by (3.40) and (3.41):

$$
\begin{aligned}
& (g) \cdot[(h)]=g \cdot(i d+i d \times(g) \cdot[(h)]) \cdot h \\
& \equiv \quad\{(g() \cdot[(h)] \text { abbreviated to } f \text { and instantiating } h \text { and } g\} \\
& f=[\underline{1}, m u l] \cdot(i d+i d \times f) \cdot(!+\langle i d, p r e d\rangle) \cdot\left(={ }_{0} ?\right) \\
& \equiv \quad\{+ \text {-functor (2.44) and identity }\} \\
& f=[\underline{1}, \text { mul }] \cdot(!+(i d \times f) \cdot\langle i d, \text { pred }\rangle) \cdot\left(==_{0} ?\right) \\
& \equiv \quad\{\times \text {-absorption (2.27) and identity }\} \\
& f=[\underline{1}, m u l] \cdot(!+\langle i d, f \cdot \operatorname{pred}\rangle) \cdot\left(=_{0} ?\right) \\
& \equiv \quad\{+ \text {-absorption (2.43) and constant } \underline{1} \text { (2.15) }\} \\
& f=[\underline{1}, m u l \cdot\langle i d, f \cdot \text { pred }\rangle] \cdot\left(={ }_{0} ?\right) \\
& \equiv \quad\{\text { McCarthy conditional (2.70) }\} \\
& f=\left(={ }_{0} ?\right) \rightarrow \underline{1}, m u l \cdot\langle i d, f \cdot \operatorname{pred}\rangle
\end{aligned}
$$

Going pointwise, we get — via (2.70) -

$$
\begin{aligned}
f 0 & =[\underline{1}, \text { mul } \cdot\langle i d, f \cdot \text { pred }\rangle]\left(i_{1} 0\right) \\
& =\quad\{+ \text { cancellation (2.40) }\} \\
& =\underline{10} \quad\{\text { constant function (2.12) }\}
\end{aligned}
$$

and

$$
f(n+1)=[\underline{1}, \text { mul } \cdot\langle i d, f \cdot \text { pred }\rangle]\left(i_{2}(n+1)\right)
$$

$$
\begin{aligned}
= & \{+ \text {-cancellation (2.40) }\} \\
= & m u l \cdot\langle i d, f \cdot \operatorname{pred}\rangle(n+1) \\
= & \{\text { pointwise definitions of split, identity, predecessor and } m u l\} \\
& (n+1) \times f n
\end{aligned}
$$

In summary, $f$ is but the well-known factorial function:

$$
\left\{\begin{array}{l}
f 0=1 \\
f(n+1)=(n+1) \times f n
\end{array}\right.
$$

This result comes to no surprise if we look at diagram (3.46) for the particular $g$ and $h$ we have considered above and recall a popular "definition" of factorial:

$$
\begin{equation*}
n!=\underbrace{n \times(n-1) \times \ldots \times 1}_{n \text { times }} \tag{3.48}
\end{equation*}
$$

In fact, $[(h)] n$ produces T-list

$$
\operatorname{Cons}(n, \operatorname{Cons}(n-1, \ldots \operatorname{Cons}(1, N i l)))
$$

as an intermediate data-structure which is consumed by $(g)$, the effect of which is but the "replacement" of Cons by $\times$ and Nil by 1, therefore accomplishing (3.48) and realizing the computation of factorial.

The moral of this example is that a function as simple as factorial can be decomposed into two components (producer/consumer functions) which share a common intermediate inductive datatype. The producer function is an anamorphism which "represents" or produces a "view" of its input argument as a value of the intermediate datatype. The consumer function is a catamorphism which reduces this intermediate data-structure and produces the final result. Like factorial, many functions can be handsomely expressed by a $(g) \cdot[(h)]$ composition for a suitable choice of the intermediate type, and of $g$ and $h$. The intermediate datastructure is said to be virtual in the sense that it only exists as a means to induce the associated pattern of recursion and disappears by calculation.

The composition $(g) \cdot[(h)]$ of a T-catamorphism with a T-anamorphism is called a T-hylomorphism ${ }^{8}$ and is denoted by $\llbracket g, h \rrbracket$. Because $g$ and $h$ fully determine the behaviour of the $\llbracket g, h \rrbracket$ function, they can be regarded as the "genes" of

[^5]the function they define. As we shall see, this analogy with biology will prove specially useful for algorithm analysis and classification.

Exercise 3.7. A way of computing $n^{2}$, the square of a given natural number $n$, is to sum up the $n$ first odd numbers. In fact, $1^{2}=1,2^{2}=1+3,3^{2}=1+3+5$, etc., $n^{2}=(2 n-1)+(n-1)^{2}$. Following this hint, express function

$$
\begin{equation*}
\operatorname{sq} n \stackrel{\text { def }}{=} n^{2} \tag{3.49}
\end{equation*}
$$

as a T-hylomorphism and encode it in HASKELL.

Exercise 3.8. Write function $x^{n}$ as a T-hylomorphism and encode it in HASKELL.

Exercise 3.9. The following function in HASKELL computes the T -sequence of all even numbers less or equal to $n$ :
$f n=$ if $n \leqslant 1$ then Nil else Cons $(m, f(m-2))$
where $m=$ if even $n$ then $n$ else $n-1$
Find its "genetic material", that is, function $g$ such that $f=[(g)]$ in


### 3.7 Inductive types more generally

So far we have focussed our attention exclusively to a particular inductive type T (3.31) - that of finite sequences of non-negative integers. This is, of course, of
a very limited scope. First, because one could think of finite sequences of other datatypes, e.g. Booleans or many others. Second, because other datatypes such as trees, hash-tables etc. exist which our notation and method should be able to take into account.

Although a generic theory of arbitrary datatypes requires a theoretical elaboration which cannot be explained at once, we can move a step further by taking the two observations above as starting points. We shall start from the latter in order to talk generically about inductive types. Then we introduce parameterization and functorial behaviour.

Suppose that, as a mere notational convention, we abbreviate every expression of the form " $1+\mathbb{N}_{0} \times \ldots$ " occurring in the previous section by " $F \ldots$..", e.g. $1+$ $\mathbb{N}_{0} \times B$ by $\mathrm{F} B$, e.g. $1+\mathbb{N}_{0} \times \mathrm{T}$ by F T

etc. This is the same as introducing a datatype-level operator

$$
\begin{equation*}
\mathrm{F} X \stackrel{\text { def }}{=} 1+\mathbb{N}_{0} \times X \tag{3.51}
\end{equation*}
$$

which maps every datatype $A$ into datatype $1+\mathbb{N}_{0} \times A$. Operator F captures the pattern of recursion which is associated to so-called "right" lists (of nonnegative integers), that is, lists which grow to the right. The slightly different pattern $\mathrm{G} X \stackrel{\text { def }}{=} 1+X \times \mathbb{N}_{0}$ will generate a different, although related, inductive type

$$
\begin{equation*}
X \cong 1+X \times \mathbb{N}_{0} \tag{3.52}
\end{equation*}
$$

- that of so-called "left" lists (of non-negative integers). And it is not difficult to think of the pattern which is merges both right and left lists and gives rise to bi-linear lists, better known as binary trees:

$$
\begin{equation*}
X \cong 1+X \times \mathbb{N}_{0} \times X \tag{3.53}
\end{equation*}
$$

One may think of many other expressions $\mathrm{F} X$ and guess the inductive datatype they generate, for instance $\mathrm{H} X \stackrel{\text { def }}{=} \mathbb{N}_{0}+\mathbb{N}_{0} \times X$ generating non-empty lists of non-negative integers $\left(\mathbb{N}_{0}^{+}\right)$. The general rule is that, given an inductive datatype definition of the form

$$
\begin{equation*}
X \cong \mathrm{~F} X \tag{3.54}
\end{equation*}
$$

(also called a domain equation), its pattern of recursion is captured by a so-called functor F .

### 3.8 Functors

The concept of a functor F, borrowed from category theory, is a most generic and useful device in programming ${ }^{9}$. As we have seen, F can be regarded as a datatype constructor which, given datatype $A$, builds a more elaborate datatype $\mathrm{F} A$; given another datatype $B$, builds a similarly elaborate datatype $\mathrm{F} B$; and so on. But what is more important and has the most beneficial consequences is that, if $F$ is regarded as a functor, then its data-structuring effect extends smoothly to functions in the following way: suppose that $B \leftarrow^{f} A$ is a function which observes $A$ into $B$, which are parameters of $\mathrm{F} A$ and $\mathrm{F} B$, respectively. By definition, if F is a functor then $\mathrm{F} B \stackrel{\mathrm{~F} f}{\rightleftarrows} \mathrm{~F} A$ exists for every such $f$ :

$\mathrm{F} f$ extends $f$ to F -structures and will, by definition, obey to two very basic properties: it commutes with identity

$$
\begin{equation*}
\mathrm{F} i d_{A}=i d_{(\mathrm{F} A)} \tag{3.55}
\end{equation*}
$$

and with composition

$$
\begin{equation*}
\mathrm{F}(g \cdot h)=(\mathrm{F} g) \cdot(\mathrm{F} h) \tag{3.56}
\end{equation*}
$$

Two simple examples of a functor follow:

- Identity functor: define $\mathrm{F} X=X$, for every datatype $X$, and $\mathrm{F} f=f$. Properties (3.55) and (3.56) hold trivially just by removing symbol F wherever it occurs.
- Constant functors: for a given $C$, define $\mathrm{F} X=C$ (for all datatypes $X$ ) and $\mathrm{F} f=i d_{C}$, as expressed in the following diagram:


Properties (3.55) and (3.56) hold trivially again.

[^6]| Data construction | Universal construct | Functor | Description |
| :---: | :---: | :---: | :--- |
| $A \times B$ | $\langle f, g\rangle$ | $f \times g$ | Product |
| $A+B$ | $[f, g]$ | $f+g$ | Coproduct |
| $B^{A}$ | $\bar{f}$ | $f^{A}$ | Exponential |

Table 3.1: Datatype constructions and associated operators.

In the same way functions can be unary, binary, etc., we can have functors with more than one argument. So we get binary functors (also called bifunctors), ternary functors etc.. Of course, properties (3.55) and (3.56) have to hold for every parameter of an $n$-ary functor. For a binary functor $\mathbf{B}$, for instance, equation (3.55) becomes

$$
\begin{equation*}
\mathrm{B}\left(i d_{A}, i d_{B}\right)=i d_{\mathrm{B}(A, B)} \tag{3.57}
\end{equation*}
$$

and equation (3.56) becomes

$$
\begin{equation*}
\mathrm{B}(g \cdot h, i \cdot j)=\mathrm{B}(g, i) \cdot \mathrm{B}(h, j) \tag{3.58}
\end{equation*}
$$

Product and coproduct are typical examples of bifunctors. In the former case one has $\mathrm{B}(A, B)=A \times B$ and $\mathrm{B}(f, g)=f \times g$ - recall (2.24). Properties (2.31) and (2.30) instantiate (3.57) and (3.58), respectively, and this explains why we called them the functorial properties of product. In the latter case, one has $\mathrm{B}(A, B)=A+B$ and $\mathrm{B}(f, g)=f+g$ - recall (2.39) - and functorial properties (2.45) and (2.44). Finally, exponentiation is a functorial construction too: assuming $A$, one has $\mathrm{F} X \stackrel{\text { def }}{=} X^{A}$ and $\mathrm{F} f \stackrel{\text { def }}{=} \overline{f \cdot a p}$ and functorial properties (2.89) and (2.90). All this is summarized in table 3.1.

Such as functions, functors may compose with each other in the obvious way: the composition of F and G, denoted F • G, is defined by

$$
\begin{align*}
(\mathrm{F} \cdot \mathrm{G}) X & \stackrel{\text { def }}{=} \mathrm{F}(\mathrm{G} X)  \tag{3.59}\\
(\mathrm{F} \cdot \mathrm{G}) f & \stackrel{\text { def }}{=} \mathrm{F}(\mathrm{G} f) \tag{3.60}
\end{align*}
$$

### 3.9 Polynomial functors

We may put constant, product, coproduct and identity functors together to obtain so-called polynomial functors, which are described by polynomial expressions,
for instance

$$
\mathrm{F} X=1+A \times X
$$

- recall (3.17). A polynomial functor is either
- a constant functor or the identity functor, or
- the (finitary) product or coproduct (sum) of other polynomial functors, or
- the composition of other polynomial functors.

So the effect on arrows of a polynomial functor is computed in an easy and structured way, for instance:

$$
\begin{aligned}
\mathrm{F} f= & (1+A \times X) f \\
= & \{\text { sum of two functors where } A \text { is a constant and } X \text { is a variable }\} \\
& (1) f+(A \times X) f \\
= & \{\text { constant functor and product of two functors }\} \\
& i d_{1}+(A) f \times(X) f \\
= & \{\text { constant functor and identity functor }\} \\
& i d_{1}+i d_{A} \times f \\
= & \{\text { subscripts dropped for simplicity }\}
\end{aligned}
$$

So, $1+A \times f$ denotes the same as $i d_{1}+i d_{A} \times f$, or even the same as $i d+i d \times f$ if one drops the subscripts.

It should be clear at this point that what was referred to in section 2.10 as a "symbolic pattern" applicable to both datatypes and arrows is after all a functor in the mathematical sense. The fact that the same polynomial expression is used to denote both the data and the operators which structurally transform such data is of great conceptual economy and practical application. For instance, once polynomial functor (3.51) is assumed, the diagrams in (3.42) can be written as simply as


It is useful to know that, thanks to the isomorphism laws studied in chapter 2, every polynomial functor $F$ may be put into the canonical form,

$$
\begin{align*}
\mathrm{FX} & \cong C_{0}+\left(C_{1} \times X\right)+\left(C_{2} \times X^{2}\right)+\cdots+\left(C_{n} \times X^{n}\right) \\
& =\sum_{i=0}^{n} C_{i} \times X^{i} \tag{3.62}
\end{align*}
$$

and that Newton's binomial formula

$$
\begin{equation*}
(A+B)^{n} \cong \sum_{p=0}^{n}{ }^{n} C_{p} \times A^{n-p} \times B^{p} \tag{3.63}
\end{equation*}
$$

can be used in such conversions. These are performed up to isomorphism, that is to say, after the conversion one gets a different but isomorphic datatype. Consider, for instance, functor

$$
\mathrm{F} X \stackrel{\text { def }}{=} A \times(1+X)^{2}
$$

(where $A$ is a constant datatype) and check the following reasoning:

$$
\begin{aligned}
\mathrm{FX}= & A \times(1+X)^{2} \\
\cong & \{\text { law }(2.102)\} \\
& A \times((1+X) \times(1+X)) \\
\cong & \{\text { law }(2.52)\} \\
& A \times((1+X) \times 1+(1+X) \times X)) \\
\cong & \{\text { laws }(2.57),(2.33) \text { and }(2.52)\} \\
& A \times((1+X)+(1 \times X+X \times X)) \\
\cong & \quad\{\text { laws }(2.57) \text { and }(2.102)\} \\
& A \times\left((1+X)+\left(X+X^{2}\right)\right) \\
\cong & \{\text { law }(2.48)\} \\
& A \times\left(1+(X+X)+X^{2}\right) \\
\cong & \{\text { canonical form obtained via laws }(2.52) \text { and }(2.103)\} \\
& \\
& \underbrace{A}_{C_{0}}+\underbrace{A \times 2}_{C_{1}} \times X+\underbrace{A}_{C_{2}} \times X^{2}
\end{aligned}
$$

Exercise 3.10. Synthesize the isomorphism $A+A \times 2 \times X+A \times X^{2}{ }_{\nu}^{\nu} A \times\left(1+X^{2}\right)$ implicit in the above reasoning.

### 3.10 Polynomial inductive types

An inductive datatype is said to be polynomial wherever its pattern of recursion is described by a polynomial functor, that is to say, wherever $F$ in equation (3.54) is polynomial. For instance, datatype T (3.31) is polynomial $(n=1)$ and its associated polynomial functor is canonically defined with coefficients $C_{0}=1$ and $C_{1}=\mathbb{N}_{0}$. For reasons that will become apparent later on, we shall always impose $C_{0} \neq 0$ to hold in a polynomial datatype expressed in canonical form.

Polynomial types are easy to encode in Haskell wherever the associated functor is in canonical polynomial form, that is, wherever one has

$$
\begin{equation*}
\mathrm{T}_{-} \cong \underbrace{\cong}_{i n_{\mathrm{T}}} \quad \sum_{i=0}^{n} C_{i} \times \mathrm{T}^{i} \tag{3.64}
\end{equation*}
$$

Then we have

$$
i n_{\mathrm{T}} \stackrel{\text { def }}{=}\left[f_{1}, \ldots, f_{n}\right]
$$

where, for $i=1, n, f_{i}$ is an arrow of type $\mathrm{T} \leftarrow C_{i} \times \mathrm{T}^{i}$. Since $n$ is finite, one may expand exponentials according to (2.102) and encode this in Haskell as follows:

$$
\operatorname{data} \mathrm{T}=C 0|C 1(C 1, \mathrm{~T})| C 2(C 2,(\mathrm{~T}, \mathrm{~T}))|\ldots| C n(C n,(\mathrm{~T}, \ldots, \mathrm{~T}))
$$

Of course the choice of symbol Ci to realize each $f_{i}$ is arbitrary ${ }^{10}$. Several instances of polynomial inductive types (in canonical form) will be mentioned in section 3.14. Section 3.18 will address the conversion between inductive datatypes induced by so-called natural transformations.

[^7]delivering every constructor in curried form.

The concepts of catamorphism, anamorphism and hylomorphism introduced in section 3.6 can be extended to arbitrary polynomial types. We devote the following sections to explaining catamorphisms in the polynomial setting. Polynomial anamorphisms and hylomorphisms will not be dealt with until chapter 9.

### 3.11 F -algebras and F -homomorphisms

Our interest in polynomial types is basically due to the fact that, for polynomial F, equation (3.54) always has a particularly interesting solution which corresponds to our notion of a recursive datatype.

In order to explain this, we need two notions which are easy to understand: first, that of an F -algebra, which simply is any function $\alpha$ of signature $A<{ }_{\longleftarrow}^{\alpha} \mathrm{F} A$. $A$ is called the carrier of F -algebra $\alpha$ and contains the values which $\alpha$ manipulates by computing new $A$-values out of existing ones, according to the F-pattern (the "type" of the algebra). As examples, consider $[\underline{0}, a d d]$ (3.30) and $i n_{\mathrm{T}}$ (3.31), which are both algebras of type $\mathrm{F} X=1+\mathbb{N}_{0} \times X$. The type of an algebra clearly determines its form. For instance, any algebra $\alpha$ of type $\mathrm{F} X=1+X \times X$ will be of form $\left[\alpha_{1}, \alpha_{2}\right]$, where $\alpha_{1}$ is a constant and $\alpha_{2}$ is a binary operator. So monoids are algebras of this type ${ }^{11}$.

Secondly, we introduce the notion of an F-homomorphism which is but a function observing a particular F -algebra $\alpha$ into another F -algebra $\beta$ :


Clearly, $f$ can be regarded as a structural translation between $A$ and $B$, that is, $A$ and $B$ have a similar structure ${ }^{12}$. Note that - thanks to (3.55) - identity functions are always (trivial) F-homomorphisms and that - thanks to (3.56) - these homomorphisms compose, that is, the composition of two F-homomorphisms is an F-homomorphism.

[^8]
### 3.12 F-catamorphisms

An F-algebra can be epic, monic or both, that is, iso. Iso F-algebras are particularly relevant to our discussion because they describe solutions to the $X \cong \mathrm{~F} X$ equation (3.54). Moreover, for polynomial F a particular iso F -algebra always exists, which is denoted by $\mu \mathrm{F} \stackrel{i n}{\leftarrow} \mathrm{~F} \mu \mathrm{~F}$ and has special properties. First, its carrier is the smallest among the carriers of other iso F-algebras, and this is why it is denoted by $\mu \mathrm{F}-\mu$ for "minimal" ${ }^{13}$. Second, it is the so-called initial F algebra. What does this mean?

It means that, for every F-algebra $\alpha$ there exists one and only one F-homomorphism between $i n$ and $\alpha$. This unique arrow mediating $i n$ and $\alpha$ is therefore determined by $\alpha$ itself, and is called the F-catamorphism generated by $\alpha$. This construct, which was introduced in 3.6 , is in general denoted by $(|\alpha|)_{F}$ :


We will drop the F subscript in $\left(\alpha \|_{F}\right.$ wherever deducible from the context, and often call $\alpha$ the "gene" of $(\alpha \alpha)_{F}$.

As happens with splits, eithers and transposes, the uniqueness of the catamorphism construct is captured by a universal property established in the class of all F-homomorphisms:

$$
\begin{equation*}
k=(|\alpha|) \quad \Leftrightarrow \quad k \cdot i n=\alpha \cdot \mathrm{F} k \tag{3.67}
\end{equation*}
$$

According to the experience gathered from section 2.13 onwards, a few properties can be expected as consequences of (3.67). For instance, one may wonder about the "gene" of the identity catamorphism. Just let $k=i d$ in (3.67) and see what happens:

$$
\begin{aligned}
i d= & (\alpha \mid) \Leftrightarrow i d \cdot i n=\alpha \cdot \mathrm{F} \text { id } \\
=\quad & \{\text { identity }(2.10) \text { and } \mathrm{F} \text { is a functor }(3.55)\} \\
i d= & (|\alpha|) \Leftrightarrow i n=\alpha \cdot i d
\end{aligned}
$$

[^9]\[

$$
\begin{aligned}
&=\quad\{\text { identity }(2.10) \text { once again }\} \\
&=\quad i d= \\
& \quad\{\alpha \mid) \Leftrightarrow i n=\alpha \\
&\{\alpha \text { replaced by } \text { in and simplifying }\} \\
&i d i n \mid)
\end{aligned}
$$
\]

Thus one finds out that the genetic material of the identity catamorphism is the initial algebra $i n$. Which is the same as establishing the reflection property of catamorphisms:

## Cata-reflection :



In a more intuitive way, one might have observed that $(\operatorname{in} \mid)$ is, by definition of $i n$, the unique arrow mediating $\mu \mathrm{F}$ and itself. But another arrow of the same type is already known: the identity $i d_{\mu \mathrm{F}}$. So these two arrows must be the same.

Another property following immediately from (3.67), for $k=(\alpha \mid)$, is

## Cata-cancellation :

$$
\begin{equation*}
(|\alpha|) \cdot i n=\alpha \cdot \boldsymbol{F}(|\alpha|) \tag{3.69}
\end{equation*}
$$

Because $i n$ is iso, this law can be rephrased as follows

$$
\begin{equation*}
(\alpha \mid)=\alpha \cdot \mathrm{F}(|\alpha|) \cdot \text { out } \tag{3.70}
\end{equation*}
$$

where out denotes the inverse of $i n$ :


Now, let $f$ be F -homomorphism (3.65) between F -algebras $\alpha$ and $\beta$. How does it relate to $(\alpha)$ and $(\beta \mid)$ ? Note that $f \cdot(\alpha \mid)$ is an arrow mediating $\mu \mathrm{F}$ and $B$. But $B$ is the carrier of $\beta$ and $(\beta \mid$ is the unique arrow mediating $\mu \mathrm{F}$ and $B$. So the two arrows are the same:

## Cata-fusion :



Of course, this law is also a consequence of the universal property, for $k=f \cdot(\alpha \mid)$ :

$$
\begin{aligned}
& f \cdot(\alpha \mid)=(\beta \beta \mid) \Leftrightarrow(f \cdot(|\alpha|)) \cdot i n=\beta \cdot \mathbf{F}(f \cdot(|\alpha|)) \\
& \Leftrightarrow \quad\{\text { composition is associative and } \mathrm{F} \text { is a functor (3.56) }\} \\
& f \cdot((\alpha \mid) \cdot i n)=\beta \cdot(\mathrm{F} f) \cdot(\mathrm{F}(|\alpha|)) \\
& \Leftrightarrow \quad\{\text { cata-cancellation (3.69) }\} \\
& f \cdot \alpha \cdot \mathbf{F}(\alpha \mid)=\beta \cdot \mathbf{F} f \cdot \mathbf{F}(|\alpha|) \\
& \Leftarrow \quad\{\text { require } f \text { to be a F-homomorphism (3.65) }\} \\
& f \cdot \alpha \cdot \mathrm{~F}(\alpha \mid)=f \cdot \alpha \cdot \mathrm{~F}(\alpha \mid) \wedge f \cdot \alpha=\beta \cdot \mathrm{F} f \\
& \Leftrightarrow \quad\{\text { simplify }\} \\
& f \cdot \alpha=\beta \cdot \mathbf{F} f
\end{aligned}
$$

The presentation of the absorption property of catamorphisms entails the very important issue of parameterization and deserves to be treated in a separate section, as follows.

### 3.13 Parameterization and type functors

By analogy with what we have done about splits (product), eithers (coproduct) and transposes (exponential), we now look forward to identifying F-catamorphisms which exhibit functorial behaviour.

Suppose that one wishes to square all numbers which are members of lists of type $T$ (3.31). It can be checked that

$$
\begin{equation*}
([\underline{N i l}, \text { Cons } \cdot(s q \times i d)] \mid) \tag{3.72}
\end{equation*}
$$

will do this for us, where $\mathbb{N}_{0} \stackrel{s q}{\longleftarrow} \mathbb{N}_{0}$ is given by (3.49). This catamorphism, which converted to pointwise notation is nothing but function $h$ which follows

$$
\left\{\begin{array}{l}
h \text { Nil }=\text { Nil } \\
h(\operatorname{Cons}(a, l))=\operatorname{Cons}(\operatorname{sq} a, h l)
\end{array}\right.
$$

maps type T to itself. This is because $s q$ maps $\mathbb{N}_{0}$ to $\mathbb{N}_{0}$. Now suppose that, instead of $s q$, one would like to apply a given function $B \stackrel{f}{\longleftarrow} \mathbb{N}_{0}$ (for some $B$ other than $\mathbb{N}_{0}$ ) to all elements of the argument list. It is easy to see that it suffices to replace $f$ for $s q$ in (3.72). However, the output type no longer is T , but rather type $\mathrm{T}^{\prime} \cong 1+B \times \mathrm{T}^{\prime}$.

Types T and $\mathrm{T}^{\prime}$ are very close to each other. They share the same "shape" (recursive pattern) and only differ with respect to the type of elements - $\mathbb{N}_{0}$ in $T$ and $B$ in $\mathrm{T}^{\prime}$. This suggests that these two types can be regarded as instances of a more generic list datatype List

$$
\begin{equation*}
\text { List } X \underset{i n=[\text { Nil,Cons }]}{\cong} 1+X \times \text { List } X \tag{3.73}
\end{equation*}
$$

in which the type of elements $X$ is allowed to vary. Thus one has $T=\operatorname{List} \mathbb{N}_{0}$ and $\mathrm{T}^{\prime}=$ List $B$.

By inspection, it can be checked that, for every $B \stackrel{f}{\longleftarrow} A$,

$$
\begin{equation*}
([\underline{N i l}, \text { Cons } \cdot(f \times i d)]) \tag{3.74}
\end{equation*}
$$

maps List $A$ to List $B$. Moreover, for $f=i d$ one has:

$$
\begin{aligned}
& ([\underline{N i l}, \text { Cons } \cdot(i d \times i d)]) \\
= & \{\text { by the } \times \text {-functor-id property }(2.31) \text { and identity }\} \\
= & ([\underline{N i l}, \text { Cons }] D \\
& \quad\{\text { cata-reflection }(3.68)\}
\end{aligned}
$$

Therefore, by defining

$$
\text { List } f \stackrel{\text { def }}{=} 0[\underline{N i l}, \text { Cons } \cdot(f \times i d)]]
$$

what we have just seen can be written thus:

$$
\operatorname{List} i d_{A}=i d_{\text {List } A}
$$

This is nothing but law (3.55) for F replaced by List. Moreover, it will not be too difficult to check that

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

also holds - cf. (3.56). Altogether, this means that List can be regarded as a functor.

In programming terminology one says that List $X$ (the "lists of $X$ s datatype") is parametric and that, by instantiating parameter $X$, one gets ground lists such as lists of integers, booleans, etc. The illustration above deepens one's understanding of parameterization by identifying the functorial behaviour of the parametric datatype along with its parameter instantiations.

All this can be broadly generalized and leads to what is commonly known by a type functor. First of all, it should be clear that the generic format

$$
\mathrm{T} \cong \mathrm{FT}
$$

adopted so far for the definition of an inductive type is not sufficiently detailed because it does not provide a parametric view of T. For simplicity, let us suppose (for the momement) that only one parameter is identified in $T$. Then we may factor this out via type variable $X$ and write (overloading symbol T)

$$
\mathrm{T} X \cong \mathrm{~B}(X, \mathrm{~T} X)
$$

where B is called the type's base functor. Binary functor $\mathrm{B}(X, Y)$ is given this name because it is the basis of the whole inductive type definition. By instantiation of $X$ one obtains F . In the example above, $\mathrm{B}(X, Y)=1+X \times Y$ and in fact $\mathrm{F} Y=\mathrm{B}\left(\mathbb{N}_{0}, Y\right)=1+\mathbb{N}_{0} \times Y$, recall (3.51). Moreover, one has

$$
\begin{equation*}
\mathrm{F} f=\mathrm{B}(i d, f) \tag{3.75}
\end{equation*}
$$

and so every F-homomorphism can be written in terms of the base-functor of F , e.g.

$$
f \cdot \alpha=\beta \cdot \mathbf{B}(i d, f)
$$

instead of (3.65).
$\mathrm{T} X$ will be referred to as the type functor generated by B :

$$
\underbrace{\mathrm{T} X}_{\text {type functor }} \cong \underbrace{\mathrm{B}(X, \mathrm{~T} X)}_{\text {base functor }}
$$

We proceed to the description of its functorial behaviour - $\mathrm{T} f$ - for a given $B \leftarrow^{f} A$. As far as typing rules are concerned, we shall have

$$
\frac{B \leftarrow^{f} A}{\mathrm{~T} B \stackrel{\mathrm{~T} f}{\leftrightarrows} \mathrm{~T} A}
$$

So we should be able to express $\mathrm{T} f$ as a $\mathrm{B}\left(A,{ }_{-}\right)$-catamorphism $(g \mid)$ :


As we know that $i n_{\mathrm{T} B}$ is the standard constructor of values of type $\mathrm{T} B$, we may put it into the diagram too:


The catamorphism's gene $g$ will be synthesized by filling the dashed arrow in the diagram with the "obvious" $\mathrm{B}(f, i d)$, whereby one gets

$$
\begin{equation*}
\mathrm{\top} f \stackrel{\text { def }}{=}\left(i n_{\mathrm{T} B} \cdot \mathrm{~B}(f, i d)\right) \tag{3.76}
\end{equation*}
$$

and a final diagram, where $i n_{\mathrm{T}_{A}}$ is abbreviated by $i n_{A}\left(\right.$ ibid. $i n_{\mathrm{T}_{B}}$ by $i n_{B}$ ):


Next, we proceed to derive the useful law of cata-absorption

$$
\begin{equation*}
(g g) \cdot \mathbf{T} f=(g \cdot \mathbf{B}(f, i d)) \tag{3.77}
\end{equation*}
$$

as consequence of the laws studied in section 3.12. Our target is to show that, for $k=(g) \cdot \mathrm{T} f$ in (3.67), one gets $\alpha=g \cdot \mathrm{~B}(f, i d)$ :

$$
\left.\begin{array}{ll} 
& (g \mid) \cdot \mathrm{T} f=(\alpha) \\
\Leftrightarrow & \{\text { type-functor definition (3.76) }\}
\end{array}\right\}
$$

The following diagram pictures this property of catamorphisms:


It remains to show that (3.76) indeed defines a functor. This can be verified by checking properties (3.55) and (3.56) for $\mathrm{F}=\mathrm{T}$ :

- Property type-functor-id, $c f$. (3.55):

$$
\begin{aligned}
& \text { Tid } \\
& =\quad\{\text { by definition (3.76) }\} \\
& \left(i n_{B} \cdot \mathrm{~B}(i d, i d) \mid\right) \\
& =\{\mathrm{B} \text { is a bi-functor (3.57) }\} \\
& \left(i n_{B} \cdot i d \mid\right) \\
& =\quad\{\text { identity and cata-reflection (3.68) \}} \\
& \text { id }
\end{aligned}
$$

- Property type-functor, $c f$. (3.56) :

$$
\begin{aligned}
& \mathrm{T}(f \cdot g) \\
& =\{\text { by definition (3.76) }\} \\
& \left(i n_{B} \cdot \mathrm{~B}(f \cdot g, i d)\right) \\
& =\quad\{i d \cdot i d=i d \text { and } \mathrm{B} \text { is a bi-functor (3.58) }\} \\
& \left(i n_{B} \cdot \mathrm{~B}(f, i d) \cdot \mathrm{B}(g, i d) \mid\right) \\
& =\{\text { cata-absorption (3.77) }\} \\
& \left(i n_{B} \cdot \mathbf{B}(f, i d)\right) \cdot \mathbf{T} g \\
& =\{\text { again cata-absorption (3.77) }\} \\
& \left(i n_{B}\right) \cdot \mathbf{T} f \cdot \mathbf{T} g \\
& =\quad\{\text { cata-reflection (3.68) followed by identity }\} \\
& \text { T } f \cdot \mathrm{~T} g
\end{aligned}
$$

Exercise 3.11. Function length $=\left(\left[\right.\right.$ zero, succ $\left.\left.\cdot \pi_{2}\right]\right)$ counts the number of elements of a finite list. If the input list has at least one element it suffices to count the elements of its tail starting with count 1 instead of 0 :

$$
\begin{equation*}
\text { length } \cdot(a:)=\left(\left[\text { one }, \text { succ } \cdot \pi_{2}\right]\right) \tag{3.78}
\end{equation*}
$$

Prove (3.78) knowing that
length $\cdot(a:)=$ succ $\cdot$ length
follows from the definition of length. (NB: assume zero $=0$ and one ${ }_{-}=1$.)

Exercise 3.12. Function concat, extracted from Haskell's Prelude, can be defined as list catamorphism,

$$
\begin{equation*}
\text { concat }=([\text { nil }, \text { conc }]) \tag{3.79}
\end{equation*}
$$

where conc $(x, y)=x+y$, nil $=[], \mathrm{B}(f, g)=i d+f \times g, \mathrm{~F} f=\mathrm{B}(i d, f)$, and $\mathrm{T} f=\operatorname{map} f$. Prove property
length $\cdot$ concat $=$ sum $\cdot$ map length
resorting to cata-fusion and cata-absorption.

### 3.14 A catalogue of standard polynomial inductive types

The following table contains a collection of standard polynomial inductive types and associated base type bi-functors, which are in canonical form (3.64). The table contains two extra columns which may be used as bookmarks for equations (3.75) and (3.76), respectively ${ }^{14}$ :

| Description | $\mathrm{T} X$ | $\mathrm{~B}(X, Y)$ | $\mathrm{B}(i d, f)$ | $\mathrm{B}(f, i d)$ |
| :--- | :--- | :---: | :---: | :---: |
| "Right" Lists | List $X$ | $1+X \times Y$ | $i d+i d \times f$ | $i d+f \times i d$ |
| "Left" Lists | LList $X$ | $1+Y \times X$ | $i d+f \times i d$ | $i d+i d \times f$ |
| Non-empty Lists | NList $X$ | $X+X \times Y$ | $i d+i d \times f$ | $f+f \times i d$ |
| Binary Trees | BTree $X$ | $1+X \times Y^{2}$ | $i d+i d \times f^{2}$ | $i d+f \times i d$ |
| "Leaf" Trees | LTree $X$ | $X+Y^{2}$ | $i d+f^{2}$ | $f+i d$ |

[^10]All type functors T in this table are unary. In general, one may think of inductive datatypes which exhibit more than one type parameter. Should $n$ parameters be identified in T , then this will be based on an $n+1$-ary base functor $\mathrm{B}, c f$.

$$
\mathrm{T}\left(X_{1}, \ldots, X_{n}\right) \cong \mathrm{B}\left(X_{1}, \ldots, X_{n}, \mathrm{~T}\left(X_{1}, \ldots, X_{n}\right)\right)
$$

So, every $n+1$-ary polynomial functor $\mathrm{B}\left(X_{1}, \ldots, X_{n}, X_{n+1}\right)$ can be identified as the basis of an inductive $n$-ary type functor (the convention is to stick to the canonical form and reserve the last variable $X_{n+1}$ for the "recursive call"). While type bi-functors ( $n=2$ ) are often found in programming, the situation in which $n>2$ is relatively rare. For instance, the combination of leaf-trees with binarytrees in (3.81) leads to the so-called "full tree" type bi-functor

| Description | $\mathrm{T}\left(X_{1}, X_{2}\right)$ | $\mathrm{B}\left(X_{1}, X_{2}, Y\right)$ | $\mathrm{B}(i d, i d, f)$ | $\mathrm{B}(f, g, i d)$ |
| :--- | :--- | :---: | :---: | :---: |
| "Full" Trees | $\mathrm{FTree}\left(X_{1}, X_{2}\right)$ | $X_{1}+X_{2} \times Y^{2}$ | $i d+i d \times f^{2}$ | $f+g \times i d$ | (32)

As we shall see later on, these types are widely used in programming. In the actual encoding of these types in HASKELL, exponentials are normally expanded to products according to (2.102), see for instance
data BTree $a=$ Empty $\mid$ Node $(a,($ BTree $a$, BTree $a))$
Moreover, one may chose to curry the type constructors as in, e.g.
data BTree $a=$ Empty $\mid$ Node $a($ BTree $a)($ BTree $a)$
Exercise 3.13. Write as a catamorphisms

- the function which counts the number of elements of a non-empty list (type NList in (3.81)).
- the function which computes the maximum element of a binary-tree of natural numbers.

Exercise 3.14. Characterize the function which is defined by $([[], h] \mid)$ for each of the following definitions of $h$ :

$$
\begin{align*}
h\left(x,\left(y_{1}, y_{2}\right)\right) & =y_{1}+[x]+y_{2}  \tag{3.83}\\
h & =\# \cdot(\operatorname{singl} \times \#)  \tag{3.84}\\
h & =\# \cdot(+\times \text { singl }) \cdot \text { swap } \tag{3.85}
\end{align*}
$$

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assuming singl $a=[a]$. Identify in (3.81) which datatypes are involved as base functors.

Exercise 3.15. Write as a catamorphism the function which computes the frontier of a tree of type LTree (3.81), listed from left to right.

Exercise 3.16. Function

$$
\begin{aligned}
& \operatorname{mirror}(\text { Leaf } a)=\text { Leaf } a \\
& \text { mirror }(\text { Fork }(x, y))=\text { Fork }(\text { mirror } y, \text { mirror } x)
\end{aligned}
$$

which mirrors binary trees of type LTree $a=$ Leaf $a \mid$ Fork (LTree $a$, LTree $a$ ) can be defined both as a catamorphism

$$
\begin{equation*}
\text { mirror }=(\text { in } \cdot(i d+\text { swap }) \mid) \tag{3.86}
\end{equation*}
$$

and as an anamorphism

$$
\begin{equation*}
\text { mirror }=[((i d+\text { swap }) \cdot \text { out })] \tag{3.87}
\end{equation*}
$$

where out is the converse of

$$
\begin{equation*}
\text { in }=[\text { Leaf }, \text { Fork }] \tag{3.88}
\end{equation*}
$$

Show that both definitions are effectively the same, that is, complete the etc steps of the rasoning:

$$
\begin{aligned}
& \text { mirror }=(\text { in } \cdot(i d+\text { swap })) \\
& \equiv \quad\{\ldots \text { etc } \ldots\} \\
& \text { mirror }=[((i d+\mathrm{swap}) \cdot \text { out })]
\end{aligned}
$$

(Hint: recall that $\mathrm{F} f=i d+f \times f$ for this type and mind the natural property of $i d+$ swap.)

Exercise 3.17. Let parametric type T be given with base B , that is, such that $\mathrm{T} f=$ ( in • B $(f, i d) \mid)$. Define the so-called triangular combinator of T , tri $f$, as follows:

$$
\begin{equation*}
\operatorname{tri} f=(\mathbf{i n} \cdot \mathrm{B}(i d, \mathrm{~T} f) \mid) \tag{3.89}
\end{equation*}
$$

Show that the instance of this combinator for type LTree $a=$ Leaf $a \mid$ Fork (LTree $a$,LTree $a$ ) - such that in $=[$ Leaf, Fork $]$ and $\mathrm{B}(f, g)=f+g \times g$-is the following function

$$
\begin{aligned}
& \text { tri }::(a \rightarrow a) \rightarrow \text { LTree } a \rightarrow \text { LTree } a \\
& \text { tri } f(\text { Leaf } x)=\text { Leaf } x \\
& \text { tri } x\left(\text { Fork }\left(t, t^{\prime}\right)\right)=\text { Fork }\left(\text { fmap } f(\text { trif } t), \text { fmap } f\left(\text { trif } t^{\prime}\right)\right)
\end{aligned}
$$

written in Haskell syntax.

### 3.15 Functors and type functors in Haskell

The concept of a (unary) functor is provided in HASKELL in the form of a particular class exporting the fmap operator:

$$
\begin{aligned}
& \text { class Functor } f \text { where } \\
& \text { fmap }::(a \rightarrow b) \rightarrow(f a \rightarrow f b)
\end{aligned}
$$

So fmap $g$ encodes $\mathrm{F} g$ once we declare F as an instance of class Functor. The most popular use of fmap has to do with Haskell lists, as allowed by declaration

$$
\begin{aligned}
& \text { instance Functor [] where } \\
& \quad \text { fmap } f[]=[] \\
& \quad \text { fmap } f(x: x s)=f x: \text { fmap } f x s
\end{aligned}
$$

in language's Standard Prelude.
In order to encode the type functors we have seen so far we have to do the same concerning their declaration. For instance, should we write

```
instance Functor BTree
    where fmap f=cataBTree (inBTree}\cdot(id+(f\timesid))
```

concerning the binary-tree datatype of (3.81) and assuming appropriate declarations of cataBTree and inBTree, then fmap is overloaded and used across such binary-trees.

Bi-functors can be added easily by writing

```
class BiFunctor f where
```

$$
\text { bmap }::(a \rightarrow b) \rightarrow(c \rightarrow d) \rightarrow(f a c \rightarrow f b d)
$$

Exercise 3.18. Declare all datatypes in (3.81) in HASkell notation and turn them into Haskell type functors, that is, define fmap in each case.

Exercise 3.19. Declare datatype (3.82) in HASKELL notation and turn it into an instance of class BiFunctor.

### 3.16 The mutual-recursion law

The theory developed so far for building (and reasoning about) recursive functions doesn't cope with mutual recursion. As a matter of fact, the pattern of recursion of a given cata(ana,hylo)morphism involves only the recursive function being defined, even though more than once, in general, as dictated by the relevant base functor.

It turns out that rules for handling mutual recursion are surprisingly simple to calculate. As motivation, recall section 2.10 where, by mixing products with coproducts, we obtained a result - the exchange rule (2.49) - which stemmed from putting together the two universal properties of product and coproduct, (2.63) and (2.65), respectively.

The question we want to address in this section is of the same brand: what can one tell about catamorphisms which output pairs of values? By (2.63), such
catamorphisms are bound to be splits, as are the corresponding genes:


As we did for the exchange rule, we put (2.63) and the universal property of catamorphisms (3.67) against each other and calculate:

$$
\begin{aligned}
& \langle f, g\rangle=(\langle h, k\rangle \mid) \\
\equiv & \{\text { cata-universal (3.67) }\} \\
& \langle f, g\rangle \cdot \text { in }=\langle h, k\rangle \cdot \mathrm{F}\langle f, g\rangle \\
\equiv & \{\times \text {-fusion (2.26) twice }\} \\
& \langle f \cdot i n, g \cdot i n\rangle=\langle h \cdot \mathrm{~F}\langle f, g\rangle, k \cdot \mathrm{~F}\langle f, g\rangle\rangle \\
\equiv & \left\{\begin{array}{l}
(2.64)\}
\end{array}\right. \\
& f \cdot i n=h \cdot \mathrm{~F}\langle f, g\rangle \wedge g \cdot i n=k \cdot \mathrm{~F}\langle f, g\rangle
\end{aligned}
$$

The rule thus obtained,

$$
\left\{\begin{array}{l}
f \cdot i n=h \cdot \mathrm{~F}\langle f, g\rangle  \tag{3.90}\\
g \cdot i n=k \cdot \mathrm{~F}\langle f, g\rangle
\end{array} \equiv\langle f, g\rangle=(\langle h, k\rangle)\right.
$$

is referred to as the mutual recursion law (or as "Fokkinga's law") and is useful in combining two mutually recursive functions $f$ and $g$

into a single catamorphism.
When applied from left to right, law (3.90) is surprisingly useful in optimizing recursive functions in a way which saves redundant traversals of the input inductive type $\mu_{\mathrm{F}}$. Let us take the Fibonacci function as example:

$$
\begin{aligned}
f i b 0 & =1 \\
f i b 1 & =1 \\
f i b(n+2) & =f i b(n+1)+f i b n
\end{aligned}
$$

It can be shown that $f i b$ is a hylomorphism of type LTree (3.81),

$$
\text { fib }=\llbracket c o u n t, f i b d \rrbracket
$$

for count $=[\underline{1}$, add $]$, add $(x, y)=x+y$ and fibd $n=$ if $n<2$ then $i_{1}$ Nil else $i_{2}(n-1, n-2)$. This hylo-factorization of $f i b$ tells its internal algorithmic structure: the divide step $[(f i b d)]$ builds a tree whose number of leaves is a Fibonacci number; the conquer step ( count) just counts such leaves.

There is, of course, much re-calculation in this hylomorphism. Can we improve its performance? The clue is to regard the two instances of $f i b$ in the recursive branch as mutually recursive over the natural numbers. This clue is suggested not only by $f i b$ having two base cases (so, perhaps it hides two functions) but also by the lookahead $n+2$ in the recursive clause.

We start by defining a function which reduces such a lookahead by 1 ,

$$
f n=f i b(n+1)
$$

Clearly, $f(n+1)=f i b(n+2)=f n+f i b n$ and $f 0=f i b 1=1$. Putting $f$ and $f i b$ together,

$$
\begin{aligned}
f 0 & =1 \\
f(n+1) & =f n+f i b n \\
f i b 0 & =1 \\
f i b(n+1) & =f n
\end{aligned}
$$

we obtain two mutually recursive functions over the natural numbers $\left(\mathbb{N}_{0}\right)$ which transform into pointfree equalities

$$
\begin{aligned}
f \cdot[\underline{0}, s u c] & =[\underline{1}, a d d \cdot\langle f, f i b\rangle] \\
f i b \cdot[\underline{0}, s u c] & =[\underline{1}, f]
\end{aligned}
$$

over

Reverse + -absorption (2.43) will further enable us to rewrite the above into

$$
\begin{aligned}
f \cdot i n & =[\underline{1}, a d d] \cdot \mathrm{F}\langle f, f i b\rangle \\
f i b \cdot i n & =\left[\underline{1}, \pi_{1}\right] \cdot \mathrm{F}\langle f, f i b\rangle
\end{aligned}
$$

thus bringing functor $\mathrm{F} f=i d+f$ explicit and preparing for mutual recursion removal:

$$
\left.\begin{array}{c}
\begin{array}{c}
f \cdot i n=[\underline{1}, a d d] \cdot \mathrm{F}\langle f, f i b\rangle \\
f i b \cdot i n=\left[\underline{1}, \pi_{1}\right] \cdot \mathrm{F}\langle f, f i b\rangle \\
\equiv \\
\{(3.90)\} \\
\langle f, f i b\rangle=0\left\langle[\underline{1}, a d d],\left[\underline{1}, \pi_{1}\right]\right\rangle D
\end{array} \\
\equiv \quad \begin{array}{c}
\{\text { exchange law }(2.49)\} \\
\langle f, f i b\rangle=0\left[\langle\underline{1}, \underline{1}\rangle,\left\langle a d d, \pi_{1}\right\rangle\right] D
\end{array} \\
\equiv \quad\left\{\text { going pointwise and denoting }\left\langle f, \text { fib } b \text { by } f i b^{\prime}\right\}\right.
\end{array}\right\} \begin{aligned}
& f i b^{\prime} 0=(1,1) \\
& f i b^{\prime}(n+1)=(x+y, x) \text { where }(x, y)=f i b^{\prime} n
\end{aligned}
$$

Since $f i b=\pi_{2} \cdot f i b^{\prime}$ we easily recover $f i b$ from $f i b^{\prime}$ and obtain the intended linear version of Fibonacci, below encoded in Haskell:

$$
\begin{aligned}
& \text { fib } n=m \text { where } \\
& \quad(-, m)=f i b^{\prime} n \\
& f i b^{\prime} 0=(1,1) \\
& \text { fib' }(n+1)=(x+y, x) \text { where }(x, y)=f i b^{\prime} n
\end{aligned}
$$

This version of $f i b$ is actually the semantics of the "for-loop" - recall (3.7) one would write in an imperative language which would initialize two global variables $x, y:=1,1$, loop over assignment $x, y:=x+y, x$ and yield the result in $y$. In the C programming language, one would write

```
int fib(int n)
{
int x=1; int y=1; int i;
for (i=1;i<=n;i++) {int a=x; x=x+y; y=a;}
return y;
};
```

where the extra variable $a$ is required for ensuring that simultaneous assignment $x, y:=x+y, x$ takes place in a sequential way.

Recall from section 3.1 that all $\mathbb{N}_{0}$ catamorphisms are of shape $([\underline{k}, g])$ and such that $([\underline{k}, g]) n=g^{n} k$, where $g^{n}$ is the $n$-th iteration of $g$, that is, $g^{0}=i d$ and
$g^{n+1}=g \cdot g^{n}$. That is, $g$ is the body of a "for-loop" which repeats itself $n$-times, starting with initial value $k$. Recall also that the for-loop combinator is nothing but the "fold combinator" (3.5) associated to the natural number data type.

In a sense, the mutual recursion law gives us a hint on how global variables "are born" in computer programs, out of the maths definitions themselves. Quite often more that two such variables are required in linearizing hylomorphisms by mutual recursion. Let us see an example. The question is: how many squares can one draw on a $n \times n$-tiled wall? The answer is given by function

$$
n s n \stackrel{\text { def }}{=} \sum_{i=1, n} i^{2}
$$

that is,

$$
\begin{aligned}
n s 0 & =0 \\
n s(n+1) & =(n+1)^{2}+n s n
\end{aligned}
$$

in Haskell. However, this hylomorphism is inefficient because each iteration involves another hylomorphism computing square numbers.

One way of improving $n s$ is to introduce function $b n m n \stackrel{\text { def }}{=}(n+1)^{2}$ and express this over (3.91),

$$
\begin{aligned}
\operatorname{bnm} 0 & =1 \\
\operatorname{bnm}(n+1) & =2 n+3+\operatorname{bnm} n
\end{aligned}
$$

hoping to blend $n s$ with $b n m$ using the mutual recursion law. However, the same problem arises in $b n m$ itself, which now depends on term $2 n+3$. We invent lin $n \stackrel{\text { def }}{=} 2 n+3$ and repeat the process, thus obtaining:

$$
\begin{aligned}
\operatorname{lin} 0 & =3 \\
\operatorname{lin}(n+1) & =2+\operatorname{lin} n
\end{aligned}
$$

By redefining

$$
\begin{aligned}
b n m^{\prime} 0 & =1 \\
b n m^{\prime}(n+1) & =\text { lin } n+b n m^{\prime} n \\
n s^{\prime} 0 & =0 \\
n s^{\prime}(n+1) & =b n m^{\prime} n+n s^{\prime} n
\end{aligned}
$$

we obtain three functions - $n s^{\prime}, b n m^{\prime}$ and $\operatorname{lin}$ - mutually recursive over the polynomial base $\mathrm{F} g=i d+g$ of the natural numbers.

Exercise 3.22 below shows how to extend (3.90) to three mutually recursive functions (3.92). (From this it is easy to extend it further to the $n$-ary case.) It is routine work to show that, by application of (3.92) to the above three functions, one obtains the linear version of $n s$ which follows:

$$
\begin{aligned}
& n s^{\prime \prime} n=a \text { where } \\
& \quad\left(a,_{-},-\right)=\text {aux } n \\
& \quad \text { aux } 0=(0,1,3) \\
& \quad \text { aux }(n+1)=\operatorname{let}(a, b, c)=\text { aux } n \operatorname{in}(a+b, b+c, 2+c)
\end{aligned}
$$

In retrospect, note that (in general) not every system of $n$ mutually recursive functions

$$
\left\{\begin{array}{l}
f_{1}=\phi_{1}\left(f_{1}, \ldots, f_{n}\right) \\
\vdots \\
f_{n}=\phi_{n}\left(f_{1}, \ldots, f_{n}\right)
\end{array}\right.
$$

involving $n$ functions and $n$ functional combinators $\phi_{1}, \ldots, \phi_{n}$ can be handled by a suitably extended version of (3.90). This only happens if all $f_{i}$ have the same "shape", that is, if they share the same base functor F.

Exercise 3.20. Use the mutual recursion law (3.90) to show that each of the two functions

$$
\left\{\begin{array} { l } 
{ \text { odd } 0 = \text { False } } \\
{ \text { odd } ( n + 1 ) = \text { even } n }
\end{array} \quad \left\{\begin{array}{l}
\text { even } 0=\text { True } \\
\text { even }(n+1)=\text { odd } n
\end{array}\right.\right.
$$

checking natural number parity can be expressed as a projection of for swap (False, True)

Encode this for-loop in C syntax.

Exercise 3.21. The following Haskell function computes the list of the first $n$ natural numbers in reverse order:

$$
\begin{aligned}
& \operatorname{insg} 0=[] \\
& \text { insg }(n+1)=(n+1): \text { insg } n
\end{aligned}
$$

1. Show that insg can also be defined as follows:

$$
\begin{aligned}
& \text { insg } 0=[] \\
& \text { insg }(n+1)=(\text { fsuc } n): \text { insg } n \\
& \text { fsuc } 0=1 \\
& \text { fsuc }(n+1)=\text { fsuc } n+1
\end{aligned}
$$

2. Based on the mutual recursion law derive from such a definition the following version of insg encoded as a for-loop:

$$
\begin{aligned}
& \text { insg }=\pi_{2} \cdot \text { insgfor } \\
& \text { insgfor }=\text { for }\left\langle(1+) \cdot \pi_{1}, \text { cons }\right\rangle(1,[])
\end{aligned}
$$

where cons $(n, m)=n: m$.

Exercise 3.22. Show that law (3.90) generalizes to more than two mutually recursive functions, in this case three:

$$
\left\{\begin{array}{l}
f \cdot i n=h \cdot \mathrm{~F}\langle f,\langle g, j\rangle\rangle  \tag{3.92}\\
g \cdot i n=k \cdot \mathrm{~F}\langle f,\langle g, j\rangle\rangle \\
j \cdot i n=l \cdot \mathrm{~F}\langle f,\langle g, j\rangle\rangle
\end{array} \quad \equiv\langle f,\langle g, j\rangle\rangle=(\langle h,\langle k, l\rangle\rangle \mathrm{D}\right.
$$

Exercise 3.23. The exponential function $e^{x}: \mathbb{R} \rightarrow \mathbb{R}$ (where " $e$ " denotes Euler's number) can be defined in several ways, one being the calculation of Taylor series:

$$
\begin{equation*}
e^{x}=\sum_{n=0}^{\infty} \frac{x^{n}}{n!} \tag{3.93}
\end{equation*}
$$

The following function, in Haskell,

$$
\begin{aligned}
& \exp :: \text { Double } \rightarrow \text { Integer } \rightarrow \text { Double } \\
& \exp x 0=1 \\
& \exp x(n+1)=x \uparrow(n+1) / \text { fac }(n+1)+(\exp x n)
\end{aligned}
$$

computes an approximation of $e^{x}$, where the second parameter tells how many terms to compute. For instance, while exp $11=2.0$, exp 110 yields 2.7182818011463845 .

Function exp x n performs badly for n larger and larger: while exp 1100 runs instantaneously, exp 11000 takes around 9 seconds, exp 12000 takes circa 33 seconds, and so on.

Decompose exp into mutually recursive functions so as to apply (3.92) and obtain the following linear version,

```
exp x n = let (e,b,c)=aux x n
    in e where
        aux x 0 = (1,2,x)
        aux x (n+1)=
            let (e,s,h)=aux x n
            in (e+h,s+1,(x/s)*h)
```

which translates directly to the encoding in $C$ :

```
float exp(float x, int n)
{
    float h=x; float e=1; int s=2; int i;
    for (i=0;i<n+1;i++) {e=e+h;h=(x/s)*h;s++;}
    return e;
};
```

Exercise 3.24. Show that, for all $n \in \mathbb{N}_{0}, n=s u c^{n} 0$. Hint: use cata-reflexion (3.68).

Mutual recursion over lists. As example of application of (3.90) for $\mu_{\mathrm{F}}$ other than $\mathbb{N}_{0}$, consider the following recursive predicate which checks whether a (nonempty) list is ordered,

$$
\begin{aligned}
& \operatorname{ord}: A^{+} \rightarrow 2 \\
& \operatorname{ord}[a]=\mathrm{TRUE} \\
& \operatorname{ord}(\operatorname{cons}(a, l))=a \geqslant(\text { listMax } l) \wedge(\text { ord } l)
\end{aligned}
$$

where $\geqslant$ is assumed to be a total order on datatype $A$ and

$$
\begin{equation*}
\text { listMax }=([i d, \max ]) \tag{3.94}
\end{equation*}
$$

computes the greatest element of a given list of $A \mathrm{~s}$ :

(In the diagram, singl $a=[a]$.)
Predicate ord is not a catamorphism because of the presence of listMax $l$ in the recursive branch. However, the following diagram depicting ord

(where $\alpha(a,(m, b)) \stackrel{\text { def }}{=} a \geqslant m \wedge b$ ) suggests the possibility of using the mutual recursion law. One only has to find a way of letting listMax depend also on ord, which isn't difficult: for any $A^{+} \xrightarrow{g} B$, one has

where the extra presence of $g$ is cancelled by projection $\pi_{1}$.
For $B=2$ and $g=o r d$ we are in position to apply Fokkinga's law and obtain:

$$
\begin{aligned}
\langle\text { listMax }, \text { ord }\rangle= & \left(\left\langle\left[i d, \max \cdot\left(i d \times \pi_{1}\right)\right],[\underline{\mathrm{TRUE}}, \alpha]\right\rangle\right) \\
= & \{\text { exchange law }(2.49)\} \\
& \left(\left[\langle i d, \underline{\mathrm{TRUE}}\rangle,\left\langle\max \cdot\left(i d \times \pi_{1}\right), \alpha\right\rangle\right]\right)
\end{aligned}
$$

Of course, ord $=\pi_{2} \cdot\langle l i s t M a x$, ord $\rangle$. By denoting the above synthesized catamorphism by $a u x$, we end up with the following version of ord:
ord $l=\operatorname{let}(a, b)=a u x l$ in $b$
where

$$
\begin{aligned}
& \text { aux }: A^{+} \rightarrow A \times 2 \\
& \text { aux }[a]=(a, \text { TRUE }) \\
& \begin{array}{l}
\text { aux }(\operatorname{cons}(a, l))=\text { let } \\
\quad \text { in } \\
(m, b)=\text { aux } l \\
\max (a, m),(a>m \wedge b))
\end{array} \\
& \text { aux }: A^{+} \rightarrow A \times 2 \\
& \text { aux }[a]=(a, \text { True }) \\
& \operatorname{aux}(\operatorname{cons}(a, l))=\text { let }(m, b)=\text { aux } l \text { in }(\max (a, m),(a>m \wedge b))
\end{aligned}
$$

Exercise 3.25. What do the following Haskell functions do?

$$
\begin{aligned}
& f_{1}[]=[] \\
& f_{1}(h: t)=h:\left(f_{2} t\right) \\
& f_{2}[]=[] \\
& f_{2}(h: t)=f_{1} t
\end{aligned}
$$

Write $f=\left\langle f_{1}, f_{2}\right\rangle$ as a list catamorphism and encode $f$ back into Haskell syntax.

### 3.17 "Banana-split": a corollary of the mutual-recursion law

Let $h=i \cdot \mathrm{~F} \pi_{1}$ and $k=j \cdot \mathrm{~F} \pi_{2}$ in (3.90). Then
$f \cdot i n=\left(i \cdot \mathrm{~F} \pi_{1}\right) \cdot \mathrm{F}\langle f, g\rangle$
$\equiv \quad\{$ composition is associative and F is a functor $\}$
$f \cdot i n=i \cdot \mathbf{F}\left(\pi_{1} \cdot\langle f, g\rangle\right)$
$\equiv \quad\{$ by $\times$-cancellation (2.22) $\}$
$f \cdot i n=i \cdot \mathrm{~F} f$
$\equiv \quad\{$ by cata-cancellation $\}$
$f=(i l)$

Similarly, from $k=j \cdot \mathrm{~F} \pi_{2}$ we get

$$
g=(j j)
$$

Then, from (3.90), we get

$$
\langle(i \mid),(j j)\rangle=\left(\left\langle\left\langle i \cdot \mathrm{~F} \pi_{1}, j \cdot \mathrm{~F} \pi_{2}\right\rangle\right\rangle\right.
$$

that is

$$
\begin{equation*}
\langle(i \mid),(|j|)\rangle=\left((i \times j) \cdot\left\langle\mathrm{F} \pi_{1}, \mathrm{~F} \pi_{2}\right\rangle\right) \tag{3.95}
\end{equation*}
$$

by (reverse) $\times$-absorption (2.27).
This law provides us with a very useful tool for "parallel loop" inter-combination: "loops" $(i)$ and $(j \mid)$ are fused together into a single "loop" $\left((i \times j) \cdot\left\langle\mathrm{F} \pi_{1}, \mathrm{~F} \pi_{2}\right\rangle\right)$. The need for this kind of calculation arises very often. Consider, for instance, the function which computes the average of a non-empty list of natural numbers,

$$
\begin{equation*}
\text { average } \stackrel{\text { def }}{=}(/) \cdot\langle\text { sum, length }\rangle \tag{3.96}
\end{equation*}
$$

where sum and length are the expected $\mathbb{N}^{+}$catamorphisms:

$$
\begin{aligned}
& \text { sum }=([i d,+]) \\
& \text { length }=\left(\left[1, \text { succ } \cdot \pi_{2}\right]\right)
\end{aligned}
$$

As defined by (3.96), function average performs two independent traversals of the argument list before division (/) takes place. Banana-split will fuse such two traversals into a single one (see function $a u x$ below), thus leading to a function which will run "twice as fast":

$$
\begin{align*}
\text { average } l=\begin{array}{l}
x / y \\
\text { where }
\end{array} & (x, y)=\text { aux } l \\
& \begin{array}{l}
\text { aux }[a]=(a, 1)
\end{array} \\
& \operatorname{aux}(\operatorname{cons}(a, l))=(a+x, y+1)  \tag{3.97}\\
& \\
& \text { where }(x, y)=\text { aux } l
\end{align*}
$$

Exercise 3.26. Calculate (3.97) from (3.96). Which of these two versions of the same function is easier to understand?

Exercise 3.27. The following diagram depicts "banana-split" (3.95):


Identify all functions $f_{1}$ to $f_{7}$.

Exercise 3.28. Show that the standard Haskell function
unzip $x s=\left(\operatorname{map} \pi_{1} x s, \operatorname{map} \pi_{2} x s\right)$
can be defined as a catamorphism (fold) thanks to (3.95). Generalize this calculation to the generic unzip function over an inductive (polynomial) type T :

$$
u n z i p_{\mathrm{T}}=\left\langle\mathrm{T} \pi_{1}, \mathrm{~T} \pi_{2}\right\rangle
$$

Suggestion: recall (3.76).

### 3.18 Inductive datatype isomorphism

## not yet available

### 3.19 Bibliography notes

It is often the case that the expressive power of a particular programming language or paradigm is counter-productive in the sense that too much freedom is given to programmers. Sooner or later, these will end up writing unintelligible (authorship dependent) code which will become a burden to whom has to maintain it. Such has been the case of imperative programming in the past (inc. assembly code), where the unrestricted use of goto instructions eventually gave place to if-then-else, while and repeat structured programming constructs.

A similar trend has been observed over the last decades at a higher programming level: arbitrary recursion and/or (side) effects have been considered harmful in functional programming. Instead, programmers have been invited to structure their code around generic program devices such as eg. fold/unfold combinators, which bring discipline to recursion. One witnesses progress in the sense that the loss of freedom is balanced by the increase of formal semantics and the availability of program calculi.

Such disciplined programming combinators have been extended from listprocessing to other inductive structures thanks to one of the most significant advances in programming theory over the last decade: the so-called functorial approach to datatypes which originated mainly from [27], was popularized by [26] and reached textbook format in [6]. A comfortable basis for exploiting polymorphism [41], the "datatypes as functors" moto has proved beneficial at a higher level of abstraction, giving birth to polytypism [22].

The literature on anas, catas and hylos is vast (see eg. [29], [21], [13]) and it is part of a broader discipline which has become known as the mathematics of program construction [2]. This chapter provides an introduction to such as discipline. Only the calculus of catamorphisms is presented. The corresponding theory of anamorphisms and hylomorphisms demands further mathematical machinery (functions generalized to binary relations) and won't be dealt with before chapters 10 and 9 . The results on mutual recursion presented in this chapter, pionered by Maarten Fokkinga [10], have been extended towards probabilistic functions [33]. They have also shown to help in program understanding and reverse engineering [40]. Recently, the whole theory has undergone significant advances throught further use of category theory notions such as adjunctions ${ }^{15}$ and conjugate functors [16, 17].

[^11]
[^0]:    ${ }^{1}$ This is nothing but a re-statement of the well-known Peano axioms for the natural numbers. Giuseppe Peano (1858-1932) was a famous Italian mathematician.
    ${ }^{2}$ Note how the singularity of type 1 ensures out a function: what would the outcome of out 0 be should $A$ be arbitrary?

[^1]:    ${ }^{3}$ See eg. section 3.6.
    ${ }^{4}$ A generalization of this property will be derived in section 3.12.

[^2]:    ${ }^{5}$ The initial and terminal datatypes do have inverses - 0 is its own "additive inverse" and 1 is its own "multiplicative inverse" - but not all the others.

[^3]:    ${ }^{6}$ By analogy, we may regard (3.14) as a "recursive definition" of number 2.

[^4]:    ${ }^{7}$ The theory which supports the statements of this paragraph will not be dealt with until chapter 9.

[^5]:    ${ }^{8}$ This terminology is derived from the Greek word $v \lambda o \sigma$ (hylos) meaning "matter".

[^6]:    ${ }^{9}$ The category theory practitioner must be warned of the fact that the word functor is used here in a too restrictive way. A proper (generic) definition of a functor will be provided later in this book.

[^7]:    ${ }^{10}$ A more traditional (but less close to (3.64)) encoding will be

    $$
    \text { data } \mathrm{T}=C 0|C 1 C 1 \mathrm{~T}| C 2 C 2 \mathrm{~T} \mathrm{~T}|\ldots| C n C n \mathrm{~T} \ldots \mathrm{~T}
    $$

[^8]:    ${ }^{11}$ But not every algebra of this type is a monoid, since the type of an algebra only fixes its syntax and does not impose any properties such as associativity, etc.
    ${ }^{12} \mathrm{Cf}$. homomorphism $=$ homo (the same) + morphos (structure, shape).

[^9]:    ${ }^{13} \mu \mathrm{~F}$ means the least fixpoint solution of equation $X \cong \mathrm{~F} X$, as will be described in chapter 9 .

[^10]:    ${ }^{14}$ Since $\left(i d_{A}\right)^{2}=i d_{\left(A^{2}\right)}$ one writes $i d^{2}$ for $i d$ in this table.

[^11]:    ${ }^{15}$ See chapter 4.

