Generic Haskell: applications

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Abstract. Generic Haskell is an extension of Haskell that supports the construction of generic programs. These lecture notes discuss three advanced generic programming applications: generic dictionaries, compressing XML documents, and the zipper: a data structure used to represent a tree together with a subtree that is the focus of attention, where that focus may move left, right, up or down the tree. When describing and implementing these examples, we will encounter some advanced features of Generic Haskell, such as type-indexed data types, dependencies between and generic abstractions of generic functions, adjusting a generic function using a default case, and generic functions with a special case for a particular constructor.

1 Introduction

A generic (or polytypic, type-indexed) function is a function that can be instantiated on many data types to obtain data type specific functionality. Examples of generic functions are the functions that can be derived in Haskell [45], such as *show*, *read*, and '=='. In the first part of these lecture notes, entitled Generic Haskell: practice and theory [22], we have introduced generic functions, and we have shown how to implement them in Generic Haskell [9]. We assume the reader is familiar with the first part of these notes. For an introduction to a related but different kind of generic programming, see Backhouse et al [2].

Why is generic programming important? Generic programming makes programs easier to write:

- Programs that could only be written in an untyped style can now be written in a language with types.
- Some programs come for free.
- Some programs are simple adjustments of library functions, instead of complicated traversal functions.

Of course not all programs become simpler when you write your programs in a generic programming language, but, on the other hand, no programs become more complicated. In this paper we will try to give you a feeling about where and when generic programs are useful.

These lecture notes describe three advanced generic programming applications: generic dictionaries, compressing XML documents, and the zipper. The applications are described in more detail below. In the examples, we will encounter several new generic programming concepts:

- Type-indexed data types. A type-indexed data type is constructed in a generic
 way from an argument data type [23]. It is the equivalent of a type-indexed
 function on the level of data types.
- Default cases. To define a generic function that is the same as another function except for a few cases we use a default case [10]. If the new definition does not provide a certain case, then the default case applies and copies the case from another function.
- Constructor cases. A constructor case of a generic program deals with a constructor of a data type that requires special treatment [10]. Constructor cases are especially useful when dealing with data types with a large number of constructors, and only a small number of constructors need special treatment.
- Dependencies and generic abstractions. To write a generic function that uses another generic function we can use a dependency or a generic abstraction [10].

We will introduce these concepts as we go along.

Example 1: Digital searching. A digital search tree or trie is a search tree scheme that employs the structure of search keys to organize information. Searching is useful for various data types, so we would like to allow for keys and information of any data type. This means that we have to construct a new kind of trie for each key type. For example, consider the data type String defined by

```
data String = Nil \mid Cons Char String.
```

We can represent string-indexed tries with associated values of type v as follows:

Such a trie for strings would typically be used for an index on texts. The constructor $Null_String$ represents the empty trie. The first component of the constructor $Node_String$ contains the value associated with Nil. The second component of $Node_String$ is derived from the constructor Cons:: Char \rightarrow String \rightarrow String. We assume that a suitable data structure, FMapChar, and an associated lookup function lookupChar:: $\forall v$. Char \rightarrow FMapChar v \rightarrow Maybe v for characters are

predefined. Given these prerequisites we can define a look-up function for strings as follows:

```
\begin{array}{ll} lookup\_String :: \mathsf{String} \to \mathsf{FMap\_String} \ \mathsf{v} \to \mathsf{Maybe} \ \mathsf{v} \\ lookup\_String \ s \ Null\_String &= Nothing \\ lookup\_String \ Nil \ (Node\_String \ tn \ tc) &= tn \\ lookup\_String \ (Cons \ c \ s) \ (Node\_String \ tn \ tc) \\ &= (lookup\_Char \ c \diamondsuit lookup\_String \ s) \ tc. \end{array}
```

To look up a non-empty string, $Cons\ c\ s$, we look up c in the FMapChar obtaining a trie, which is then recursively searched for s. Since the look-up functions have result type Maybe v, we use the reverse monadic composition of the Maybe monad, denoted by ' \diamondsuit ', to compose $lookup_String$ and lookupChar.

```
(\diamondsuit) :: (a \rightarrow Maybe b) \rightarrow (b \rightarrow Maybe c) \rightarrow a \rightarrow Maybe c
(f \diamondsuit g) a = \mathbf{case} \ f \ a \ \mathbf{of} \ \{Nothing \rightarrow Nothing; Just \ b \rightarrow g \ b\}
```

Consider now the data type Bush of binary trees with characters in the leaves:

```
\mathbf{data} \; \mathsf{Bush} = \mathit{Leaf} \; \mathsf{Char} \; | \; \mathit{Bin} \; \mathsf{Bush} \; \mathsf{Bush}.
```

Bush-indexed tries can be represented by the following data type:

```
\begin{aligned} \mathbf{data} \ \mathsf{FMap\_Bush} \ \mathsf{v} &= \mathit{Null\_Bush} \\ &\mid \mathit{Node\_Bush} \ (\mathsf{FMapChar} \ \mathsf{v}) \\ &\quad (\mathsf{FMap\_Bush} \ (\mathsf{FMap\_Bush} \ \mathsf{v})). \end{aligned}
```

Again, we have two components, one to store values constructed by Leaf, and one for values constructed by Bin. The corresponding look-up function is given by

```
lookup\_Bush :: Bush \rightarrow FMap\_Bush \ v \rightarrow Maybe \ v
lookup\_Bush \ b \ Null\_Bush = Nothing
lookup\_Bush \ (Leaf \ c) \ (Node\_Bush \ tl \ tf) = lookupChar \ c \ tl
lookup\_Bush \ (Bin \ bl \ br) \ (Node\_Bush \ tl \ tf)
= (lookup\_Bush \ bl \ \diamondsuit \ lookup\_Bush \ br) \ tf.
```

One can easily recognize that not only the look-up functions, but also the data types for the tries are instances of an underlying generic pattern. In the following section we will show how to define a trie and associated functions generically for arbitrary data types.

Example 2: Compressing XML documents. The extensible markup language XML [49] is used to mark up text with structure information. XML documents may become (very) large because of the markup that is added to the content. A good XML compressor can compress an XML document by quite a large factor.

An XML document is usually structured according to a DTD (Document Type Definition), a specification that describes which tags may be used in the XML document, and in which positions and order they have to be. A DTD is, in a way, the *type* of an XML document. An XML document is called *valid* with respect to a certain DTD if it follows the structure that is specified by that DTD. An XML compressor can use information from the DTD to obtain better compression. For example, consider the following small XML file:

```
<book lang="English">
<title> Dead Famous </title>
<author> Ben Elton </author>
<date> 2001 </date>
</book>.
```

This file may be compressed by separating the structure from the data, and compressing the two parts separately. For compressing the structure we can make good use of the DTD of the document.

```
<!ELEMENT book (title,author,date,(chapter)*)>
<!ELEMENT title (#PCDATA)>
<!ELEMENT author (#PCDATA)>
<!ELEMENT date (#PCDATA)>
<!ELEMENT chapter (#PCDATA)>
<!ATTLIST book lang (English | Dutch) #REQUIRED>
```

If we know how many elements and attributes, say n, appear in the DTD (the DTD above document contains 6 elements), we can replace the markup of an element in an XML file which is valid with respect to the DTD by a natural number between 0 and n-1, or by log_2 n bits. This is one of the main ideas behind XMill [36]. We improve on XMill by only recording the choices made in an XML document. In the above document, there is a choice for the language of the book, and the number of chapters it has. All the other elements are not encoded, since they can be inferred from the DTD. Section 3 describes a tool based on this idea, which was first described by Jansson and Jeuring in the context of data conversion [26, 30]. We use HaXml [51] to translate a DTD to a data type, and we construct generic functions for separating the contents (the strings) and the shape (the constructors) of a value of a data type, and for encoding the shape of a value of a data type using information about the (number of) constructors of the data type.

XML compressors are just one class of XML tools that are easily implemented as generic programs. Other XML tools that can be implemented as generic programs are XML editors, XML databases, and XML version management tools.

Example 3: Zipper. The zipper [24] is a data structure that is used to represent a tree together with a subtree that is the focus of attention, where that focus may move left, right, up, or down the tree. For example, the zipper corresponding to the data type Bush, called Loc_Bush, is defined by

```
 \begin{aligned} \textbf{type} \ \mathsf{Loc\_Bush} &= (\mathsf{Bush}, \mathsf{Context\_Bush}) \\ \textbf{data} \ \mathsf{Context\_Bush} &= \mathit{Top} \\ \mid \mathit{BinL} \ \mathsf{Context\_Bush} \ \mathsf{Bush} \\ \mid \mathit{BinR} \ \mathsf{Bush} \ \mathsf{Context\_Bush}. \end{aligned}
```

Using the type of locations we can efficiently navigate through a tree. For example:

```
\begin{array}{lll} \textit{down\_Bush} & & \text{:: Loc\_Bush} \rightarrow \textit{Loc\_Bush} \\ \textit{down\_Bush} \; (\textit{Leaf} \; a, \, c) & = (\textit{Leaf} \; a, \, c) \\ \textit{down\_Bush} \; (\textit{Bin} \; tl \; tr, \, c) & = (tl, \textit{BinL} \; c \; tr) \\ \textit{right\_Bush} & & \text{:: Loc\_Bush} \rightarrow \textit{Loc\_Bush} \\ \textit{right\_Bush} \; (tl, \textit{BinL} \; c \; tr) & = (tr, \textit{BinR} \; tl \; c) \\ \textit{right\_Bush} \; m & = m. \end{array}
```

The navigation function down_Bush moves the focus of attention to the leftmost subtree of the current node; right_Bush moves the focus to its right sibling.

Huet [24] defines the zipper data structure for rose trees and for the data type Bush, and gives the generic construction in words. In Section 4 we describe the zipper in more detail and show how to define a zipper for an arbitrary data type.

Other applications of generic programming. Besides the applications mentioned in the examples above, there are several application areas in which generic programming can be used.

- Haskell's deriving construct. Haskell's deriving construct is used to generate code for for example the equality function, and for functions for reading and showing values of data types. Only the classes Eq. Ord, Enum, Bounded, Show and Read can be derived. The definitions of (equivalents of) the derived functions can be found in the library of Generic Haskell.
- Compiler functions. Several functions that are used in compilers are generic functions: garbage collectors, tracers, debuggers, test tools [7, 34], etc.
- Typed term processing. Functions like pattern matching, term rewriting and unification are generic functions, and have been implemented as generic functions in [31, 28, 29].

The form and functionality of these applications is exactly determined by the structure of the input data.

Maybe the most common applications of generic programming can be found in functions that traverse data built from rich mutually-recursive data types with many constructors, and which perform computations on a single (or a couple of) constructor(s). For example, consider a function which traverses an abstract syntax tree and returns the free variables in the tree. Only for the variable constructor something special has to be done, in all other cases the variables collected at the children have to be passed on to the parent. This function can be defined as an instance of a Generic Haskell library function *crush* [41], together with a special constructor case for variables [10]. For more examples of generic traversals, see Lämmel and Peyton Jones [35].

The Generic Haskell code for the programs discussed in these lecture notes can be downloaded from the applications page on http://www.generic-haskell.org/.

On notation. To improve readability, the notation for generic functions we use in these notes slightly differs from the notation used in the first part of these notes [22]. For example, in the first part of these lecture notes we write:

```
equal { | Char |}
                                               = eqChar
equal{Int}
                                               = eqInt
equal { | Unit | Unit | Unit |
                                               = True
equal\{:+:\}\ equal\{(Inl\ a)\ (Inl\ a')\}
                                               = eqa \ a \ a'
                                               = False
equal\{:+:\}\ equal\{(Inl\ a)\ (Inr\ b')\}
equal\{:+:\}\ equal\{(Inr\ b)\ (Inl\ a')\}
                                               = False
equal\{:+:\}\ equal\{(Inr\ b)\ (Inr\ b')\}
                                               = eqb \ b \ b'
equal\{:*:\}\ eqa\ eqb\ (a:*:b)\ (a':*:b') = eqa\ a\ a' \land eqb\ b\ b'
equal\{|Con c|\} eqa (Con a) (Con b)
                                               = eqa \ a \ b.
```

The function equal is a generic function which recurses over the type structure of the argument type. The recursion is implicit in the arguments equ and eqb in the :+: and :*: cases. In this part of the lecture notes we will instead write:

```
equal { | Char|}
                                           = eqChar
equal { | Int | }
                                           = eqInt
equal {| Unit | Unit | Unit |
                                           = True
equal\{|a:+:b|\}\ (Inl\ a)\ (Inl\ a')
                                           = equal\{a\} \ a \ a'
equal\{|a:+:b|\}\ (Inl\ a)\ (Inr\ b')
                                           = False
equal\{|a:+:b|\}\ (Inr\ b)\ (Inl\ a')
                                           = False
equal\{a:+:b\}\ (Inr\ b)\ (Inr\ b')
                                           = equal\{\{b\}\} b b'
equal\{|a:*:b|\}\ (a:*:b)\ (a':*:b') = equal\{|a|\}\ a\ a' \land equal\{|b|\}\ b\ b'
equal\{|Con\ c\ a|\}\ (Con\ a)\ (Con\ b) = equal\{|a|\}\ a\ b.
```

Here the recursion over the type structure is explicit: $equal\{\{a:*:b\}\}$ is expressed in terms of $equal\{\{a\}\}$ and $equal\{\{b\}\}$. We think this style is more readable, especially when a generic function depends on another generic function. Functions written in the latter style can be translated to the former style and vice versa, so no expressiveness is lost or gained; the only difference is readability. The Generic Haskell compiler does not accept explicit recursive functions yet, but probably will do so in the near future. A formal description of Generic Haskell with explicit recursion is given by Löh et al. [37].

Organization. The rest of this paper is organized as follows. Section 2 introduces generic dictionaries, and implements them in Generic Haskell. Section 3 describes XCOMPREZ, a compressor for XML documents. Section 4 develops a generic zipper data structure. Section 5 summarizes the main points and concludes.

These lecture notes contain exercises. Solutions to the exercises can be found on the webpage for the Generic Haskell project: www.generic-haskell.org.

2 Generic dictionaries

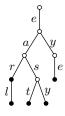
A trie is a search tree scheme that employs the structure of search keys to organize information. Tries were originally devised as a means to represent a

collection of records indexed by strings over a fixed alphabet. Based on work by Wadsworth and others, Connelly et al. [11] generalized the concept to permit indexing by elements built according to an arbitrary signature. In this section we go one step further and define tries and operations on tries generically for arbitrary data types of arbitrary kinds, including parameterized and nested data types. The material in this section is largely taken from [18].

2.1 Introduction

The concept of a trie was introduced by Thue in 1912 as a means to represent a set of strings, see [33]. In its simplest form a trie is a multiway branching tree

where each edge is labelled with a character. For example, the set of strings { ear, earl, east, easy, eye} is represented by the trie depicted on the right. Searching in a trie starts at the root and proceeds by traversing the edge that matches the first character, then traversing the edge that matches the second character, and so forth. The search key is a member of the represented set if the search stops in a node that is marked—marked nodes are drawn as filled circles on the right. Tries can also be used to represent finite maps. In this case marked nodes additionally contain values associated with the strings. Interestingly, the move from sets to finite maps is not a mere variation of the scheme. As we shall see it is essential for the further development.



On a more abstract level a trie itself can be seen as a composition of finite maps. Each collection of edges descending from the same node constitutes a finite map sending a character to a trie. With this interpretation in mind it is relatively straightforward to devise an implementation of string-indexed tries. If strings are defined by the following data type:

$$data String = Nil \mid Cons Char String,$$

we can represent string-indexed tries with associated values of type v as follows.

Here, *Null_String* represents the empty trie. The first component of the constructor *Node_String* contains the value associated with *Nil*. Its type is Maybe v instead of v since *Nil* may not be in the domain of the finite map represented by the trie. In this case the first component equals *Nothing*. The second component corresponds to the edge map. To keep the introductory example manageable we

implement FMapChar using ordered association lists.

```
 \begin{array}{ll} \textbf{type} \ \mathsf{FMapChar} \ \mathsf{v} = [(\mathsf{Char}, \mathsf{v})] \\ lookupChar & :: \ \forall \mathsf{v} \cdot \mathsf{Char} \to \mathsf{FMapChar} \ \mathsf{v} \to \mathsf{Maybe} \ \mathsf{v} \\ lookupChar \ c \ [] & = Nothing \\ lookupChar \ c \ ((c', v) : x) \\ | \ c < c' & = Nothing \\ | \ c = c' & = Just \ v \\ | \ c > c' & = lookupChar \ c \ x \\ \end{array}
```

Note that *lookupChar* has result type Maybe v. If the key is not in the domain of the finite map, *Nothing* is returned.

Building upon lookupChar we can define a look-up function for strings. To look up the empty string we access the first component of the trie. To look up a non-empty string, say, $Cons\ c\ s$ we look up c in the edge map obtaining a trie, which is then recursively searched for s.

```
\begin{array}{lll} lookup\_String :: \forall v \,.\, String \rightarrow \mathsf{FMap\_String} \,\, v \rightarrow \mathsf{Maybe} \,\, v \\ lookup\_String \,\, s \,\, Null\_String &= Nothing \\ lookup\_String \,\, Nil \,\, (Node\_String \,\, tn \,\, tc) &= tn \\ lookup\_String \,\, (Cons \,\, c \,\, s) \,\, (Node\_String \,\, tn \,\, tc) &= \\ (lookup\_Char \,\, c \,\, \diamond \,\, lookup\_String \,\, s) \,\, tc \end{array}
```

In the last equation we use monadic composition to take care of the error signal *Nothing*.

Based on work by Wadsworth and others, Connelly et al. [11] have generalized the concept of a trie to permit indexing by elements built according to an arbitrary signature, that is, by elements of an arbitrary non-parameterized data type. The definition of $lookup_String$ already gives a clue what a suitable generalization might look like: the trie $Node_String$ tn tc contains a finite map for each constructor of the data type String; to look up Cons c s the look-up functions for the components, c and s, are composed. Generally, if we have a data type with k constructors, the corresponding trie has k components. To look up a constructor with n fields, we must select the corresponding finite map and compose n look-up functions of the appropriate types. If a constructor has no fields (such as Nil), we extract the associated value.

As a second example, consider the data type of external search trees:

```
data Dict = Tip String \mid Node Dict String Dict.
```

A trie for external search trees represents a finite map from Dict to some value type v. It is an element of $FMap_Dict\ v$ given by

```
\begin{aligned} \textbf{data} \ \mathsf{FMap\_Dict} \ \mathsf{v} &= \mathit{Null\_Dict} \\ &\mid \mathit{Node\_Dict} \ (\mathsf{FMap\_String} \ \mathsf{v}) \\ &\quad (\mathsf{FMap\_Dict} \ (\mathsf{FMap\_String} \ (\mathsf{FMap\_Dict} \ \mathsf{v}))). \end{aligned}
```

The data type FMap_Dict is a nested data type, since the recursive call on the right hand side, FMap_Dict (FMap_String (FMap_Dict v)), is a substitution instance of the left hand side. Consequently, the look-up function on external search trees requires polymorphic recursion.

```
\begin{array}{ll} lookup\_Dict :: \forall \texttt{v} \,.\, \mathsf{Dict} \to \mathsf{FMap\_Dict} \,\, \mathsf{v} \to \mathsf{Maybe} \,\, \mathsf{v} \\ lookup\_Dict \,\, d \,\, Null\_Dict &= Nothing \\ lookup\_Dict \,\, (Tip \,\, s) \,\, (Node\_Dict \,\, tl \,\, tn) &= lookup\_String \,\, s \,\, tl \\ lookup\_Dict \,\, (Node \,\, m \,\, s \,\, r) \,\, (Node\_Dict \,\, tl \,\, tn) &= \\ (lookup\_Dict \,\, m \,\, \diamondsuit \,\, lookup\_String \,\, s \,\, \diamondsuit \,\, lookup\_Dict \,\, r) \,\, tn \end{array}
```

Looking up a node involves two recursive calls. The first, $lookup_Dict\ m$, is of type $Dict \to FMap_Dict\ X \to Maybe\ X$ where $X = FMap_String\ (FMap_Dict\ v)$, which is a substitution instance of the declared type.

Note that it is absolutely necessary that $\mathsf{FMap_Dict}$ and $lookup_Dict$ are parametric with respect to the codomain of the finite maps. If we restrict the type of $lookup_Dict$ to $\mathsf{Dict} \to \mathsf{FMap_Dict} \ \mathsf{T} \to \mathsf{Maybe} \ \mathsf{T}$ for some fixed type T , the definition no longer type-checks. This also explains why the construction does not work for the finite set abstraction.

Generalized tries make a particularly interesting application of generic programming. The central insight is that a trie can be considered as a *type-indexed data type*. This renders it possible to define tries and operations on tries generically for arbitrary data types. We already have the necessary prerequisites at hand: we know how to define tries for sums and for products. A trie for a sum is essentially a product of tries and a trie for a product is a composition of tries. The extension to arbitrary data types is then uniquely defined. Mathematically speaking, generalized tries are based on the following isomorphisms.

$$\begin{array}{ll} 1 \rightarrow_{\operatorname{fin}} v & \cong v \\ (t_1 + t_2) \rightarrow_{\operatorname{fin}} v \cong (t_1 \rightarrow_{\operatorname{fin}} v) \times (t_2 \rightarrow_{\operatorname{fin}} v) \\ (t_1 \times t_2) \rightarrow_{\operatorname{fin}} v \cong t_1 \rightarrow_{\operatorname{fin}} (t_2 \rightarrow_{\operatorname{fin}} v) \end{array}$$

Here, $t \to_{\text{fin}} v$ denotes the set of all finite maps from t to v. Note that $t \to_{\text{fin}} v$ is sometimes written $v^{[t]}$, which explains why these equations are also known as the 'laws of exponentials'.

2.2 Signature

To put the above idea in concrete terms we will define a type-indexed data type FMap , which has the following kind for types t of kind \star .

$$\mathsf{FMap}\{\mathsf{l}\mathsf{t}::\star\}::\star\to\star$$

So FMap assigns a type constructor of kind $\star \to \star$ to each key type t of kind \star .

We will implement the following operations on tries.

```
\begin{split} & empty\{\!\!\mid t\} \quad :: \forall v \cdot \mathsf{FMap}\{\!\!\mid t\} \mid v \\ & isempty\{\!\!\mid t\} \mid :: \forall v \cdot \mathsf{FMap}\{\!\!\mid t\} \mid v \to \mathsf{Bool} \\ & single\{\!\!\mid t\} \quad :: \forall v \cdot (t,v) \to \mathsf{FMap}\{\!\!\mid t\} \mid v \\ & lookup\{\!\!\mid t\} \quad :: \forall v \cdot t \to \mathsf{FMap}\{\!\!\mid t\} \mid v \to \mathsf{Maybe} \mid v \\ & insert\{\!\!\mid t\} \quad :: \forall v \cdot (v \to v \to v) \to (t,v) \to (\mathsf{FMap}\{\!\!\mid t\} \mid v \to \mathsf{FMap}\{\!\!\mid t\} \mid v) \\ & merge\{\!\!\mid t\} \quad :: \forall v \cdot (v \to v \to v) \to (\mathsf{FMap}\{\!\!\mid t\} \mid v \to \mathsf{FMap}\{\!\!\mid t\} \mid v) \\ & delete\{\!\!\mid t\} \quad :: \forall v \cdot t \to (\mathsf{FMap}\{\!\!\mid t\} \mid v \to \mathsf{FMap}\{\!\!\mid t\} \mid v) \end{split}
```

The value $empty\{t\}$ is the empty trie. The function $isempty\{t\}$ takes a trie and determines whether or not it is empty. The function $single\{t\}$ (t,v) constructs a trie that contains the binding (t,v) as its only element. The function $lookup\{t\}$ takes a key and a trie and looks up the value associated with the key. The function $insert\{t\}$ inserts a new binding into a trie, and $merge\{t\}$ combines two tries. The function $delete\{t\}$ takes a key and a trie, and removes the binding for the key from the trie. The two functions $insert\{t\}$ and $merge\{t\}$ take as a first argument a so-called $combining\ function$, which is applied whenever two bindings have the same key. For instance, $\lambda new\ old\ \rightarrow\ new$ is used as the combining function for $insert\{t\}$ if the new binding is to override an old binding with the same key. For finite maps of type $\mathsf{FMap}\{t\}$ Int addition may also be a sensible choice. Interestingly, we will see that the combining function is not only a convenient feature for the user; it is also necessary for defining $insert\{t\}$ and $merge\{t\}$ generically for all types!

2.3 Properties

Each of the operations specified in the previous section satisfies a number of laws, which hold generically for all instances of t. These properties formally specify parts of the informal descriptions of the operations given there, and can be proved for the definitions given in the following sections using fixed point induction. See Hinze [19, 21] for examples of proofs of properties of generic functions.

```
\begin{aligned} &lookup\{\{\mathsf{t}\}\ k\ (empty\{\{\mathsf{t}\}\}) = Nothing \\ &lookup\{\{\mathsf{t}\}\ k\ (single\{\{\mathsf{t}\}\ (k_1,v_1)) = \mathbf{if}\ k =: k_1\ \mathbf{then}\ Just\ v_1\ \mathbf{else}\ Nothing \\ &lookup\{\{\mathsf{t}\}\ k\ (merge\{\{\mathsf{t}\}\ c\ t_1\ t_2) = combine\ c\ (lookup\{\{\mathsf{t}\}\ k\ t_1)\ (lookup\{\{\mathsf{t}\}\ k\ t_2), \\ \end{aligned}
```

where *combine* combines two values of type Maybe:

```
combine :: \forall v \cdot (v \rightarrow v \rightarrow v) \rightarrow (Maybe \ v \rightarrow Maybe \ v) combine c Nothing Nothing = Nothing combine c Nothing (Just v_2) = Just v_2 combine c (Just v_1) Nothing = Just v_1 combine c (Just v_1) (Just v_2) = Just (c v_1 \ v_2).
```

The last law, for instance, states that looking up a key in the merge of two tries yields the same result as looking up the key in each trie separately and then

combining the results. If the combining form c is associative,

$$c \ v_1 \ (c \ v_2 \ v_3) = c \ (c \ v_1 \ v_2) \ v_3,$$

then $merge\{\{t\}\}\ c$ is associative, as well. Furthermore, $empty\{\{t\}\}\ is$ the left and the right unit of $merge\{\{t\}\}\ c$:

```
merge\{\{t\}\}\ c\ (empty\{\{t\}\})\ x=x merge\{\{t\}\}\ c\ x\ (empty\{\{t\}\})=x merge\{\{t\}\}\ c\ x_1\ (merge\{\{t\}\}\ c\ x_2\ x_3)=merge\{\{t\}\ c\ (merge\{\{t\}\}\ c\ x_1\ x_2)\ x_3.
```

The functions insert and delete satisfy the following laws: for all k, v, and c,

$$\begin{array}{l} insert \{ \mid \!\!\! t \mid \!\!\! \} \ c \ (k,v) \ (empty \{ \mid \!\!\! t \mid \!\!\! \}) = single \{ \mid \!\!\! t \mid \!\!\! \} \ (k,v) \\ delete \{ \mid \!\!\! t \mid \!\!\! \} \ k \ (single \{ \mid \!\!\! t \mid \!\!\! \} \ (k,v)) = empty \{ \mid \!\!\! t \mid \!\!\! \}. \\ \end{array}$$

The operations satisfy many more laws, but we will omit them here.

2.4 Type-indexed tries

We have already noted that generalized tries are based on the laws of exponentials.

$$\begin{array}{lll} 1 \mathop{\rightarrow_{\mathrm{fin}}} v & \cong v \\ (t_1 + t_2) \mathop{\rightarrow_{\mathrm{fin}}} v \cong (t_1 \mathop{\rightarrow_{\mathrm{fin}}} v) \times (t_2 \mathop{\rightarrow_{\mathrm{fin}}} v) \\ (t_1 \times t_2) \mathop{\rightarrow_{\mathrm{fin}}} v \cong t_1 \mathop{\rightarrow_{\mathrm{fin}}} (t_2 \mathop{\rightarrow_{\mathrm{fin}}} v) \end{array}$$

In order to define the notion of finite map it is customary to assume that each value type v contains a distinguished element or base point \bot_v , see [11]. A finite map is then a function whose value is \bot_v for all but finitely many arguments. For the implementation of tries it is, however, inconvenient to make such a strong assumption (though one could use type classes for this purpose).

Instead, we explicitly add a base point when necessary motivating the following definition of FMap, our first example of a type-indexed data type.

```
\begin{split} &\mathsf{FMap}\{\mathsf{l}\, :: \, \star \} & :: \, \star \to \, \star \\ &\mathsf{FMap}\{\mathsf{Unit}\} \, \, \mathsf{v} &= \mathsf{Maybe} \, \mathsf{v} \\ &\mathsf{FMap}\{\mathsf{Int}\} \, \, \mathsf{v} &= \mathit{Patricia}.\mathsf{Dict} \, \mathsf{v} \\ &\mathsf{FMap}\{\mathsf{Char}\} \, \, \mathsf{v} &= \mathsf{FMapChar} \, \mathsf{v} \\ &\mathsf{FMap}\{\mathsf{t}_1 :+: \, \mathsf{t}_2\} \, \, \mathsf{v} = \mathsf{FMap}\{\mathsf{t}_1\} \, \, \mathsf{v} \, \times_{\bullet} \, \mathsf{FMap}\{\mathsf{t}_2\} \, \, \mathsf{v} \\ &\mathsf{FMap}\{\mathsf{t}_1 :+: \, \mathsf{t}_2\} \, \, \mathsf{v} = \mathsf{FMap}\{\mathsf{t}_1\} \, \, (\mathsf{FMap}\{\mathsf{t}_2\} \, \, \mathsf{v}) \\ &\mathsf{FMap}\{\mathsf{Con} \, \, \mathsf{t}\} \, \, \mathsf{v} &= \mathsf{FMap}\{\mathsf{t}\} \, \, \mathsf{v} \end{split}
```

Here, (\times_{\bullet}) is the type of optional pairs.

$$\mathbf{data} \ \mathsf{a} \times_{\bullet} \mathsf{b} = Null \mid Pair \ \mathsf{a} \ \mathsf{b}$$

Instead of optional pairs we could also use ordinary pairs in the definition of FMap:

$$\mathsf{FMap}\{t_1 : + : t_2\}\ \mathsf{v} = \mathsf{FMap}\{t_1\}\ \mathsf{v} : * : \mathsf{FMap}\{t_2\}\ \mathsf{v}.$$

This representation has, however, two major drawbacks: (i) it relies in an essential way on lazy evaluation and (ii) it is inefficient, see [18].

We assume there exists a suitable library implementing finite maps with integer keys. Such a library could be based, for instance, on a data structure known as a *Patricia tree* [44]. This data structure fits particularly well in the current setting since Patricia trees are a variety of tries. For clarity, we will use qualified names when referring to entities defined in the hypothetical module *Patricia*.

A few remarks are in order. FMap is a type-indexed data type [23]. The only way to construct values of type FMap is by means of the functions in the interface.

Furthermore, in contrast with type-indexed functions, the constructor index Con doesn't mention a constructor description anymore. This is because a type cannot depend on a value, so the constructor description can never be used in the definition of a type-indexed data type.

Since the trie for the unit type is given by Maybe v rather than v itself, tries for isomorphic types are, in general, not isomorphic. We have, for instance, Unit \cong Unit:*:Unit (ignoring laziness) but FMap{Unit} v = Maybe v \ncong Maybe (Maybe v) = FMap{Unit:*:Unit} v. The trie type Maybe (Maybe v) has two different representations of the empty trie: Nothing and Just Nothing. However, only the first one will be used in our implementation. Similarly, Maybe v \times Maybe v has two elements, Null and Pair Nothing Nothing, that represent the empty trie. Again, only the first one will be used.

As mentioned in Section 2.2, the kind of FMap for types of kind \star is $\star \to \star$. For type constructors with higher-order kinds, the kind of FMap looks surprisingly similar to the type of type-indexed functions for higher-order kinds. A trie on the type List a is a trie for the type List, applied to a trie for the type a:

$$\mathsf{FMap}\{\{f::\star\to\star\}\}::(\star\to\star)\to(\star\to\star).$$

The 'type' of a type-indexed type is a kind-indexed kind. In general, we have:

```
\begin{split} \mathsf{FMap} \{ \mathsf{f} :: \kappa \} & :: \mathsf{FMAP} \{ \!\! [\kappa] \!\! \} \; \mathsf{f} \\ \mathsf{FMAP} \{ \!\! [\kappa :: \square] \!\! \} & :: \square \\ \mathsf{FMAP} \{ \!\! [\star] \!\! \} & = \star \to \star \\ \mathsf{FMAP} \{ \!\! [\kappa \to \nu] \!\! \} & = \mathsf{FMAP} \{ \!\! [\kappa] \!\! \} \to \mathsf{FMAP} \{ \!\! [\nu] \!\! \}, \end{split}
```

where the box \square is the type of a kind, a so-called *super-kind*.

Example 1. Let us specialize FMap to the following data types.

```
 \begin{aligned} \mathbf{data} \ \mathsf{List} \ \mathsf{a} &= Nil \mid \mathit{Cons} \ \mathsf{a} \ (\mathsf{List} \ \mathsf{a}) \\ \mathbf{data} \ \mathsf{Tree} \ \mathsf{a} \ \mathsf{b} &= \mathit{Tip} \ \mathsf{a} \mid \mathit{Node} \ (\mathsf{Tree} \ \mathsf{a} \ \mathsf{b}) \ \mathsf{b} \ (\mathsf{Tree} \ \mathsf{a} \ \mathsf{b}) \\ \mathbf{data} \ \mathsf{Fork} \ \mathsf{a} &= \mathit{Fork} \ \mathsf{a} \ \mathsf{a} \\ \mathbf{data} \ \mathsf{Sequ} \ \mathsf{a} &= \mathit{EndS} \mid \mathit{ZeroS} \ (\mathsf{Sequ} \ (\mathsf{Fork} \ \mathsf{a})) \mid \mathit{OneS} \ \mathsf{a} \ (\mathsf{Sequ} \ (\mathsf{Fork} \ \mathsf{a})) \end{aligned}
```

These types are represented by (see also section 3.1 in the first part of these lecture notes [22]):

```
List = Fix (\Lambda List . \Lambda a . Unit :+: a :*: List a)

Tree = Fix (\Lambda Tree . \Lambda a b . a :+: Tree a b :*: b :*: Tree a b)

Fork = \Lambda a . a :*: a

Sequ = Fix (\Lambda Sequ . \Lambda a . Unit :+: Sequ (Fork a) :+: a :*: Sequ (Fork a)).
```

Recall that (:*:) binds stronger than (:+:). Consequently, the corresponding trie types are

```
\begin{split} \mathsf{FMap\_List} &= \mathsf{Fix} \; ( \Lambda \mathsf{FMap\_List} \; . \, \Lambda \mathsf{fa} \; . \, \mathsf{Maybe} \; \times_{\bullet} \; \mathsf{fa} \; \cdot \; \mathsf{FMap\_List} \; \mathsf{fa}) \\ \mathsf{FMap\_Tree} &= \mathsf{Fix} \; ( \Lambda \mathsf{FMap\_Tree} \; . \, \Lambda \mathsf{fa} \; \mathsf{fb} \; . \\ & \qquad \qquad \mathsf{fa} \; \times_{\bullet} \\ & \qquad \qquad \mathsf{FMap\_Tree} \; \mathsf{fa} \; \mathsf{fb} \; \cdot \; \mathsf{fb} \; \cdot \; \mathsf{FMap\_Tree} \; \mathsf{fa} \; \mathsf{fb}) \\ \mathsf{FMap\_Fork} &= \Lambda \mathsf{fa} \; . \; \mathsf{fa} \; \cdot \; \mathsf{fa} \\ \mathsf{FMap\_Sequ} &= \mathsf{Fix} \; ( \Lambda \mathsf{FMap\_Sequ} \; . \, \Lambda \mathsf{fa} \; . \\ & \qquad \qquad \mathsf{Maybe} \; \times_{\bullet} \\ & \qquad \qquad \mathsf{FMap\_Sequ} \; ( \mathsf{FMap\_Fork} \; \mathsf{fa}) \; \times_{\bullet} \\ & \qquad \qquad \mathsf{fa} \; \cdot \; \mathsf{FMap\_Sequ} \; ( \mathsf{FMap\_Fork} \; \mathsf{fa}) ). \end{split}
```

As an aside, note that we interpret $a \times_{\bullet} b \times_{\bullet} c$ as the type of optional triples and not as nested optional pairs:

```
data a \times_{\bullet} b \times_{\bullet} c = Null \mid Triple a b c.
```

Now, since Haskell permits the definition of higher-order kinded data types, the second-order type constructors above can be directly coded as data types. All we have to do is to bring the equations into an applicative form.

These types are the parametric variants of FMap_String and FMap_Dict defined in Section 2.1: we have FMap_String \approx FMap_List FMapChar (corresponding to String \approx List Char) and FMap_Dict \approx FMap_Tree FMap_String FMap_String (corresponding to Dict \approx Tree String String). Things become interesting if we consider nested data types.

```
 \begin{aligned} \textbf{data} \; & \mathsf{FMap\_Fork} \; \mathsf{fa} \; \mathsf{v} \; = Node\_Fork \; (\mathsf{fa} \; \mathsf{(fa} \; \mathsf{v})) \\ \textbf{data} \; & \mathsf{FMap\_Sequ} \; \mathsf{fa} \; \mathsf{v} \; = Null\_Sequ \\ & | \; Node\_Sequ \; (\mathsf{Maybe} \; \mathsf{v}) \\ & \; (\mathsf{FMap\_Sequ} \; (\mathsf{FMap\_Fork} \; \mathsf{fa}) \; \mathsf{v}) \\ & \; (\mathsf{fa} \; (\mathsf{FMap\_Sequ} \; (\mathsf{FMap\_Fork} \; \mathsf{fa}) \; \mathsf{v})) \end{aligned}
```

The generalized trie of a nested data type is a second-order nested data type! A nest is termed second-order, if a parameter that is instantiated in a recursive call ranges over type constructors of first-order kind. The trie FMap_Sequ is a second-order nest since the parameter fa of kind $\star \to \star$ is changed in the recursive calls. By contrast, FMap_Tree is a first-order nest since its instantiated parameter v has kind \star . It is quite easy to produce generalized tries that are both first- and second-order nests. If we swap the components of Sequ's third constructor—OneS a (Sequ (Fork a)) becomes OneS (Sequ (Fork a)) a—then the third component of FMap_Sequ has type FMap_Sequ (FMap_Fork fa) (fa v) and since both fa and v are instantiated, FMap_Sequ is consequently both a first- and a second-order nest.

2.5 Empty tries

The empty trie is defined as follows.

```
type Empty{[★]} t
                            = \forall v . \mathsf{FMap}\{t\} v
type Empty\{ \kappa \to \nu \} t = \forall a. Empty\{ \kappa \} a \to Empty\{ \nu \} (t a)
                             :: Empty\{ [\kappa] \} t
empty\{|\mathbf{t}::\kappa|\}
empty{|Unit|}
                             = Nothing
empty{|Char|}
empty{Int}
                             = Patricia.empty
empty\{a:+:b\}
                             = Null
                            = empty\{|a|\}
empty\{|a:*:b|\}
empty\{\{Con \ c \ a\}\}
                             = empty\{a\}
```

The definition already illustrates several interesting aspects of programming with generalized tries. First, the explicit polymorphic type of empty is necessary to make the definition work. Consider the line $empty\{a:*:b\}$, which is of type $\forall v . \mathsf{FMap}\{a\}$ ($\mathsf{FMap}\{b\}\ v$). It is defined in terms of $empty\{a\}$, which is of type $\forall v . \mathsf{FMap}\{a\}\ v$. That means that $empty\{a\}$ is used polymorphically. In other words, empty makes use of polymorphic recursion!

Suppose we want to define a function emptyI that is almost the same as the function empty, but uses a different value, say emptyIntTrie, for the empty trie for integers. The definition of FMap says that emptyIntTrie has to be a Patricia tree, but that might be changed in the definition of FMap. Then we can use a default case [10] to define emptyI in terms of empty as follows:

```
emptyI\{t :: \kappa\} :: Empty(\kappa) t

emptyI\{lnt\} = emptyIntTrie

emptyI\{a\} = empty\{a\}.
```

So the function emptyI is equal to the function empty in all cases except for the Int case, where it uses a special empty trie.

Example 2. Let us specialize empty to lists and binary random-access lists.

```
\begin{array}{l} empty\_List :: \forall \mathsf{fa} \ . \ (\forall \mathsf{w} \ . \ \mathsf{fa} \ \mathsf{w}) \to (\forall \mathsf{v} \ . \ \mathsf{FMap\_List} \ \mathsf{fa} \ \mathsf{v}) \\ empty\_List \ ea = Null\_List \\ empty\_Fork :: \forall \mathsf{fa} \ . \ (\forall \mathsf{w} \ . \ \mathsf{fa} \ \mathsf{w}) \to (\forall \mathsf{v} \ . \ \mathsf{FMap\_Fork} \ \mathsf{fa} \ \mathsf{v}) \\ empty\_Fork \ ea = Node\_Fork \ ea \\ empty\_Sequ :: \forall \mathsf{fa} \ . \ (\forall \mathsf{w} \ . \ \mathsf{fa} \ \mathsf{w}) \to (\forall \mathsf{v} \ . \ \mathsf{FMap\_Sequ} \ \mathsf{fa} \ \mathsf{v}) \\ empty\_Sequ \ ea = Null\_Sequ \\ \end{array}
```

The second function, $empty_Fork$, illustrates the polymorphic use of the parameter: ea has type $\forall w$. fa w but is used as an element of fa (fa w). The functions $empty_List$ and $empty_Sequ$ show that the 'mechanically' generated definitions can sometimes be slightly improved: the argument ea is not needed.

The function $isempty\{t\}$ takes a trie and determines whether it is empty.

```
type lsEmpty{\{\star\}} t
                                = \forall v . \mathsf{FMap}\{|t|\} \ v \to \mathsf{Bool}
type lsEmpty{[\kappa \rightarrow \nu]} t = \forall a . lsEmpty{[\kappa]} a \rightarrow lsEmpty{[\nu]} (t a)
                                           :: IsEmpty\{ \kappa \} t
            isempty\{|\mathbf{t}::\kappa|\}
                                           = isNothing v
             isempty{\{|Unit|\}} v
             isempty{Char} m
                                           = null m
             isempty{Int} m
                                           = Patricia.isempty m
             isempty\{a:+:b\}\ Null = True
             isempty{a :+: b} d
                                           = False
             isempty\{|\mathbf{a}:*:\mathbf{b}|\}\ d
                                           = isempty\{|a|\} d
            isempty{\{|Con ca|\} d}
                                           = isempty\{a\} d
```

Function *isempty* assumes that tries are in 'normal form', so the empty trie is always represented by *Null*, and not, for example, by *Pair Null Null*.

Example 3. Let us specialize isempty to lists and binary random-access lists.

```
isempty\_List
                                                                :: \forall fa . (\forall w . fa w \rightarrow Bool) \rightarrow
                                                                           (\forall v . \mathsf{FMap\_List} \mathsf{ fa } \mathsf{ v} \to \mathsf{Bool})
isempty\_List\ iea\ Null\_List
                                                                = True
isempty_List iea (Node_List tn tc)
                                                                = False
                                                                :: \forall fa . (\forall w . fa w \rightarrow Bool) \rightarrow
isempty_Fork
                                                                           (\forall v . \mathsf{FMap}_\mathsf{F}\mathsf{ork} \; \mathsf{fa} \; \mathsf{v} \to \mathsf{Bool})
isempty_Fork iea (Node_Fork tf)
                                                                = iea tf
isempty\_Sequ
                                                                :: \forall fa . (\forall w . fa w \rightarrow Bool) \rightarrow
                                                                           (\forall v . \mathsf{FMap\_Sequ} \ \mathsf{fa} \ v \to \mathsf{Bool})
isempty_Sequ iea Null_Sequ
isempty\_Sequ iea (Node\_Sequ tv tf ts) = False
```

2.6 Singleton tries

The function $single\{t\}$ (t,v) constructs a trie that contains the binding (t,v) as its only element. To construct a trie in the sum case, we have to return a Pair, of which only one component is inhabited. The other component is the empty trie. This means that single depends on empty. Generic Haskell supports dependencies, so we can use both the empty trie and the single trie in the sum, product, and constructor cases of function single. The dependency shows in the type of the function single: on higher-order kinds, the type mentions the type of empty.

```
type Single{\{\star\} t = \forall v . (t, v) \rightarrow \mathsf{FMap}\{t\} v

type Single{\{\kappa \rightarrow \nu\} t = \forall a . \mathsf{Empty}\{\{\kappa\}\} a \rightarrow \mathsf{Single}\{\{\kappa\}\} a \rightarrow \mathsf{Single}\{\{\nu\}\} (t a)
```

Plain generic functions can be seen as catamorphisms [38, 42] over the structure of data types. With dependencies, we also get the power of paramorphisms [40].

```
\begin{array}{lll} single \{\!\{ \mathsf{t} :: \kappa \}\!\} & :: \mathsf{Single} \{\!\{ \kappa \}\!\} \ \mathsf{t} \\ single \{\!\{ \mathsf{Unit} \}\!\} \ (Unit, v) & = Just \ v \\ single \{\!\{ \mathsf{Char} \}\!\} \ (c, v) & = [(c, v)] \\ single \{\!\{ \mathsf{Int} \}\!\} \ (i, v) & = Patricia.single \ (i, v) \\ single \{\!\{ \mathsf{a} :+ : \mathsf{b} \}\!\} \ (Inl \ a, v) & = Pair \ (single \{\!\{ \mathsf{a} \}\!\} \ (a, v)) \ (empty \{\!\{ \mathsf{b} \}\!\}) \\ single \{\!\{ \mathsf{a} :+ : \mathsf{b} \}\!\} \ (a :* : b, v) & = single \{\!\{ \mathsf{a} \}\!\} \ (a, single \{\!\{ \mathsf{b} \}\!\} \ (b, v)) \\ single \{\!\{ \mathsf{Con} \ c \ a \}\!\} \ (Con \ b, v) & = single \{\!\{ \mathsf{a} \}\!\} \ (b, v) \end{array}
```

Example 4. Let us again specialize the generic function to lists and binary random-access lists.

```
single\_List :: \forall k \text{ fa } . (\forall w \text{ . fa } w) \rightarrow (\forall w \text{ . } (k, w) \rightarrow \text{ fa } w)
                             \rightarrow (\forall v. (List k, v) \rightarrow FMap\_List fa v)
single\_List\ ea\ sa\ (Nil,v)
                                                = Node\_List (Just v) ea
single\_List\ ea\ sa\ (Cons\ k\ ks, v)\ =
 Node\_List\ Nothing\ (sa\ (k, single\_List\ ea\ sa\ (ks, v)))
single\_Fork :: \forall k \text{ fa } . (\forall w . \text{ fa } w) \rightarrow (\forall w . (k, w) \rightarrow \text{ fa } w)
                               \rightarrow (\forall v. (Fork k, v) \rightarrow FMap\_Fork fa v)
single\_Fork\ ea\ sa\ (Fork\ k_1\ k_2,v) = Node\_Fork\ (sa\ (k_1,sa\ (k_2,v)))
single\_Sequ :: \forall k \text{ fa } . (\forall w . \text{ fa } w) \rightarrow (\forall w . (k, w) \rightarrow \text{ fa } w)
                               \rightarrow (\forall v . (Sequ k, v) \rightarrow FMap\_Sequ fa v)
single\_Sequ\ ea\ sa\ (EndS,v)
                                                = Node\_Sequ (Just v) Null\_Sequ ea
single\_Sequ\ ea\ sa\ (ZeroS\ s,v)
   Node_Sequ Nothing
                    (single\_Sequ\ (empty\_Fork\ ea)\ (single\_Fork\ ea\ sa)\ (s,v))
single\_Sequ\ ea\ sa\ (OneS\ k\ s,v)\ =
   Node_Sequ Nothing
                    Null\_Sequ
                    (sa\ (k, single\_Sequ\ (empty\_Fork\ ea)\ (single\_Fork\ ea\ sa)\ (s, v)))
```

Again, we can simplify the 'mechanically' generated definitions: since the definition of Fork does not involve sums, $single_Fork$ does not require its first argument, ea, which can be safely removed.

2.7 Look up

The look-up function implements the scheme discussed in Section 2.1.

```
lookup\{|\mathbf{t}::\kappa|\}
                                                :: Lookup{ [\kappa] } t
lookup{|Unit|} Unit fm
                                                = fm
                                                = lookupChar \ c \ fm
lookup\{|Char|\} \ c \ fm
lookup{Int} i fm
                                                = Patricia.lookup i fm
lookup{a :+: b} t Null
                                                = Nothing
lookup\{a:+:b\}\ (Inl\ a)\ (Pair\ fma\ fmb) = lookup\{a\}\ a\ fma
lookup\{a: +: b\} (Inr\ b) (Pair\ fma\ fmb) = lookup\{b\}\ b\ fmb
lookup\{|a:*:b|\}\ (a:*:b)\ fma
                                               = (lookup\{a\} \ a \diamondsuit lookup\{b\} \ b) fma
lookup\{|Con \ d \ a|\} \ (Con \ b) \ fm
                                                = lookup\{a\} \ b \ fm
```

On sums the look-up function selects the appropriate map; on products it 'composes' the look-up functions for the components. Since *lookup* has result type Maybe v, we use monadic composition.

Example 5. Specializing $lookup\{t\}$ to concrete instances of t is by now probably a matter of routine.

```
lookup\_List :: \forall k \text{ fa } . (\forall w . k \rightarrow \text{fa } w \rightarrow \text{Maybe } w)
                               \rightarrow (\forall v . List k \rightarrow FMap\_List fa v \rightarrow Maybe v)
lookup_List la ks Null_List
                                                                         = Nothing
lookup_List la Nil (Node_List tn tc)
                                                                         = tn
lookup_List la (Cons k ks) (Node_List tn tc)
                                                                         = (la \ k \diamond lookup\_List \ la \ ks) \ tc
lookup\_Fork :: \forall k \text{ fa } . (\forall w . k \rightarrow fa w \rightarrow Maybe w)
                                \rightarrow (\forall v . Fork k \rightarrow FMap\_Fork fa v <math>\rightarrow Maybe v)
lookup\_Fork\ la\ (Fork\ k_1\ k_2)\ (Node\_Fork\ tf)
                                                                         = (la k_1 \diamondsuit la k_2) tf
lookup\_Sequ :: \forall fa k. (\forall w.k \rightarrow fa w \rightarrow Maybe w)
                                \rightarrow (\forall v . \mathsf{Sequ} \ \mathsf{k} \rightarrow \mathsf{FMap}\_\mathsf{Sequ} \ \mathsf{fa} \ \mathsf{v} \rightarrow \mathsf{Maybe} \ \mathsf{v})
lookup_Sequ la s Null_Sequ
                                                                         = Nothing
lookup_Sequ la EndS (Node_Sequ te tz to)
                                                                         = te
lookup_Sequ la (ZeroS s) (Node_Sequ te tz to)
   lookup_Sequ (lookup_Fork la) s tz
lookup\_Sequ\ la\ (OneS\ a\ s)\ (Node\_Sequ\ te\ tz\ to) =
   (la\ a \diamondsuit lookup\_Sequ\ (lookup\_Fork\ la)\ s)\ to
```

The function $lookup_List$ generalizes $lookup_String$ defined in Section 2.1; we have $lookup_String \approx lookup_List$ lookupChar.

2.8 Inserting and merging

Insertion is defined in terms of merge and single.

Function *insert* is defined as a *generic abstraction*. A generic abstraction is a generic function that is defined in terms of another generic function. The abstracted type parameter is, however, restricted to types of a fixed kind. In the above case, *insert* only works for types of kind \star . In the exercise at the end of this section you will define *insert* as a generic function that works for type constructors of all kinds.

Merging two tries is surprisingly simple. Given the function *combine* defined in section 2.3, and a function for merging two association lists

```
\begin{array}{ll} \textit{mergeChar} & :: \ \forall v \,.\, (v \rightarrow v \rightarrow v) \rightarrow \\ & (\mathsf{FMapChar} \ v \rightarrow \mathsf{FMapChar} \ v \rightarrow \mathsf{FMapChar} \ v) \\ \textit{mergeChar} \ c \ [] \ x' = x' \\ \textit{mergeChar} \ c \ x \ [] \ = x \\ \textit{mergeChar} \ c \ ((k, v) : x) \ ((k', v') : x') \\ | \ k < k' = (k, v) : \textit{mergeChar} \ c \ x \ ((k', v') : x') \\ | \ k = k' = (k, c \ v \ v') : \textit{mergeChar} \ c \ x \ x' \\ | \ k > k' = (k', v') : \textit{mergeChar} \ c \ ((k, v) : x) \ x', \end{array}
```

we can define *merge* as follows.

```
type Merge\{\{\kappa\}\}\ t = \forall v. (v \rightarrow v \rightarrow v) \rightarrow FMap\{\{t\}\}\ v \rightarrow FMap\{\{t
```

For the sum case, we have to distinguish between empty and nonempty tries:

```
 merge \{ | \mathbf{a} : + : \mathbf{b} \} \ c \ d \ Null \\ merge \{ | \mathbf{a} : + : \mathbf{b} \} \ c \ Null \ d \\ merge \{ | \mathbf{a} : + : \mathbf{b} \} \ c \ (Pair \ x \ y) \ (Pair \ v \ w) = \\ Pair \ (merge \{ | \mathbf{a} \} \ c \ x \ v) \ (merge \{ | \mathbf{b} \} \ c \ y \ w).
```

The most interesting equation is the product case. The tries d and d' are of type $\mathsf{FMap}\{a\}$ ($\mathsf{FMap}\{b\}\ v$). To merge them we can recursively call $\mathit{merge}\{a\}$; we must, however, supply a combining function of type $\forall v \cdot \mathsf{FMap}\{b\}\ v \to \mathsf{FMap}\{b\}$

 $\mathsf{FMap}\{|\mathsf{b}|\;\mathsf{v}\to\mathsf{FMap}\{|\mathsf{b}|\;\mathsf{v}.\;\mathsf{A}\;\mathsf{moment's}\;\mathsf{reflection}\;\mathsf{reveals}\;\mathsf{that}\;\mathit{merge}\{|\mathsf{b}|\}\;c$ is the desired combining function.

```
merge\{a: *: b \mid c \ d \ d' = merge\{a\} \ (merge\{b\} \ c) \ d \ d'
```

The definition of *merge* shows that it is sometimes necessary to implement operations more general than immediately needed. If $Merge\{\![\star]\!]$ t had been the simpler type $\forall v . FMap\{\![t]\!]$ $v \to FMap\{\![t]\!]$ $v \to FMap\{\![t]\!]$ v, then we would not have been able to give a defining equation for :*:.

Example 6. To complete the picture let us again specialize the merging operation for lists and binary random-access lists. The different instances of *merge* are surprisingly concise (only the types look complicated).

```
merge\_List :: \forall fa . (\forall w . (w \rightarrow w \rightarrow w) \rightarrow (fa w \rightarrow fa w \rightarrow fa w))
                     \rightarrow (\forall v . (v \rightarrow v \rightarrow v))
                           \rightarrow FMap_List fa v \rightarrow FMap_List fa v \rightarrow FMap_List fa v)
merge_List ma c Null_List t
merge\_List\ ma\ c\ t\ Null\_List
                                                                                       = t
merge_List ma c (Node_List tn tc) (Node_List tn' tc')
   Node\_List\ (combine\ c\ tn\ tn')
                  (ma\ (merge\_List\ ma\ c)\ tc\ tc')
merge\_Fork :: \forall fa . (\forall w . (w \rightarrow w \rightarrow w) \rightarrow (fa w \rightarrow fa w \rightarrow fa w))
                      \rightarrow (\forall v \:.\: (v \to v \to v)
                            \rightarrow FMap_Fork fa v \rightarrow FMap_Fork fa v \rightarrow FMap_Fork fa v)
merge_Fork ma c (Node_Fork tf) (Node_Fork tf')
   Node\_Fork\ (ma\ (ma\ c)\ tf\ tf')
merge\_Sequ :: \forall fa . (\forall w . (w \rightarrow w \rightarrow w) \rightarrow (fa w \rightarrow fa w \rightarrow fa w))
                      \rightarrow (\forall v . (v \rightarrow v \rightarrow v)
                           \rightarrow FMap_Sequ fa v \rightarrow FMap_Sequ fa v \rightarrow FMap_Sequ fa v)
merge_Sequ ma c Null_Sequ t
                                                                                       = t
merge_Sequ ma c t Null_Sequ
merge_Sequ ma c (Node_Sequ te tz to) (Node_Sequ te' tz' to') =
   Node_Sequ (combine c te te')
                   (merge_Sequ (merge_Fork ma) c tz tz')
                   (ma (merge_Sequ (merge_Fork ma) c) to to')
```

2.9 Deleting

Function $delete\{t\}$ takes a key and a trie, and removes the binding for the key from the trie. For the Char case we need a help function that removes an element from an association list:

```
deleteChar :: \forall v . Char \rightarrow FMapChar v \rightarrow FMapChar v,
```

and similarly for the Int case. Function delete is defined as follows:

```
 \begin{array}{lll} delete\{\{\mathtt{l}::\kappa\}\} & :: \ \mathsf{Delete}\{\{\kappa\}\} \ \mathsf{t} \\ delete\{\{\mathtt{lnit}\}\} \ \mathit{Unit} \ \mathit{fm} &= \mathit{Nothing} \\ delete\{\{\mathtt{lnit}\}\} \ \mathit{c} \ \mathit{fm} &= \mathit{deleteChar} \ \mathit{c} \ \mathit{fm} \\ delete\{\{\mathtt{lnit}\}\} \ \mathit{i} \ \mathit{fm} &= \mathit{Patricia.delete} \ \mathit{i} \ \mathit{fm}. \\ \end{array}
```

All cases except the product case are straightforward. In the product case, we have to remove a binding for a product (a:*:b). We do this by using a to lookup the trie d in which there is a binding for b. Then we remove the binding for b in d, obtaining a trie d'. If d' is empty, then we delete the complete binding for a in d, otherwise we insert the binding (a, d') in the original trie d. Here we pass as a combining function $\lambda x \ y \to x$, which overwrites existing bindings in a trie. From this description it follows that the function delete depends on the functions lookup, insert (which depends on function empty), and isempty. Here we need the kind-indexed typed version of function insert, as defined in the exercise at the end of this section.

```
 \begin{tabular}{ll} type & Delete\{\![\kappa]\!] t & = \forall v \,.\, t \to \mathsf{FMap}\{\![t]\!] v \to \mathsf{FMap}\{\![t]\!] v \\ type & Delete\{\![\kappa \to \nu]\!] t = \forall a \,.\, \mathsf{Lookup}\{\![\kappa]\!] a \\ & \to \mathsf{Insert}\{\![\kappa]\!] a \\ & \to \mathsf{IsEmpty}\{\![\kappa]\!] a \\ & \to \mathsf{Delete}\{\![\kappa]\!] a \\ & \to \mathsf{Delete}\{\![\nu]\!] (t \,a) \\ \end{tabular}
```

Function *delete* should also maintain the invariant that the empty trie is represented by *Null*, and not by *Pair Null Null*, for example. It is easy to adapt the above definition such that this invariant is maintained.

Since the type of *delete* is rather complex because of the dependencies, we only give the instance of *delete* on List.

```
delete\_List :: \forall k \text{ fa .} (\forall w . k \rightarrow fa w \rightarrow Maybe w)
                              \rightarrow (\forall w. (w \rightarrow w \rightarrow w) \rightarrow (k, w) \rightarrow fa \ w \rightarrow fa \ w)
                              \rightarrow (\forallw . fa w \rightarrow Bool)
                              \rightarrow (\forall w. fa w)
                              \rightarrow (\forall w. (k \rightarrow fa w \rightarrow fa w))
                              \rightarrow (\forall v . List k \rightarrow FMap\_List fa v \rightarrow FMap\_List fa v)
delete_List la ia iea ea da Nil Null_List
                                                                                        = Null\_List
                                                                                        = Node\_List\ Nothing\ tc
delete_List la ia iea ea da Nil (Node_List tn tc)
                                                                                       = Null\_List
delete_List la ia iea ea da (Cons a as) Null_List
delete\_List\ la\ ia\ iea\ ea\ da\ (Cons\ a\ as)\ (Node\_List\ tn\ tc) =
   case (la a \diamondsuit delete\_List la ia iea ea da as) tc of
     Nothing \rightarrow tc
     Just\ tc'\mid iea\ tc' 
ightarrow da\ a\ tc
                 | otherwise \rightarrow ia (\lambda x y \rightarrow x) (a, tc') tc
```

2.10 Related work

Knuth [33] attributes the idea of a trie to Thue who introduced it in a paper about strings that do not contain adjacent repeated substrings [47]. De la Briandais [13] recommended tries for computer searching. The generalization of tries from strings to elements built according to an arbitrary signature was discovered by Wadsworth [50] and others independently since. Connelly et al. [11] formalized the concept of a trie in a categorical setting: they showed that a trie is a functor and that the corresponding look-up function is a natural transformation.

The first implementation of generalized tries was given by Okasaki in his recent textbook on functional data structures [43]. Tries for parameterized types like lists or binary trees are represented as Standard ML functors. While this approach works for regular data types, it fails for nested data types such as Sequ. In the latter case data types of second-order kind are indispensable.

Exercise 1. Define function insert as a generic function with a kind-indexed kind. You can download the code for the functions described in this section from http://www.generic-haskell.org, including the solution to this exercise. You might want to avoid looking at the implementation of insert while solving this exercise.

Exercise 2. Define a function update, which updates a binding in a trie.

$$update\{|t|\} :: \forall v.(t,v) \rightarrow \mathsf{FMap}\{|t|\} v \rightarrow \mathsf{Maybe}(\mathsf{FMap}\{|t|\} v)$$

If there is no binding for the value of type t in trie of type $\mathsf{FMap}\{t\}$ v, update returns Nothing.

3 XComprez: a generic XML compressor

The extensible markup language XML is a popular standard for describing documents with markup (or structure). XML documents may become (very) large because of the markup that is added to the content. A lot of diskspace and bandwidth is used to store and send XML documents. XML compressors reduce the size of an XML document, sometimes by a considerable factor. This section describes a generic XML compressor based on the ideas described in the context of data conversion by Jansson and Jeuring [26, 30].

This section shows how an XML compressor is implemented as a generic program, and it briefly discusses which other classes of XML tools would profit from an implementation as a generic program. The example shows how Generic Haskell can be used to implement XML tools whose behaviour depends on the DTD or Schema of the input XML document. Example tools include XML editors, databases, and compressors.

Generic Haskell is ideally suited for implementing XML tools:

- Knowledge of the DTD can be used to provide more precise functionality, such as manipulations of an XML document that preserve validity in an XML editor, or better compression in an XML compressor.
- Generic Haskell programs are typed. Consequently, valid documents are transformed to valid documents, possibly structured according to another DTD. Thus Generic Haskell supports constructing type correct XML tools.
- The generic features of Generic Haskell make XML tools easier to implement in a surprisingly small amount of code.
- The Generic Haskell compiler may perform all kinds of advanced optimisations on the code, such as partial evaluation or deforestation, which are difficult to conceive or implement by an XML tool developer.

3.1 Implementing an XML compressor as a generic program

We have implemented an XML compressor, called XCOMPREZ, as a generic program. XCOMPREZ separates structure from contents, compresses the structure using knowledge about the DTD, and compresses the contents using the Unix compress utility [52]. Thus we replace each element, or rather, the pair of open and close keywords of the element, by the minimal number of bits required for the element given the DTD. We distinguish four components in the tool:

- a component that translates a DTD to a data type,
- a component that separates a value of a data type into its structure and its contents.
- a component that encodes the structure replacing constructors by bits,
- and a component for compressing the contents.

Of course, we have also implemented a decompressor, but since it is very similar to the compressor, we omit its description. See the website for XCOMPREZ [32] for the latest developments on XCOMPREZ. The Generic Haskell source code for XCOMPREZ can be obtained from the website.

Translating a DTD to a data type. A DTD can be translated to one or more Haskell data types. For example, the following DTD:

```
<!ELEMENT book
                        (title, author, date, (chapter)*)>
    <!ELEMENT title
                        (#PCDATA)>
    <!ELEMENT author (#PCDATA)>
    <!ELEMENT date
                        (#PCDATA)>
    <!ELEMENT chapter (#PCDATA)>
    <!ATTLIST book lang (English | Dutch) #REQUIRED> ,
can be translated to the following data types:
      data Book
                       = Book Book_Attrs Title Author Date [Chapter]
      data Book\_Attrs = Book\_Attrs\{bookLang :: Lang\}
                       = English \mid Dutch
      data Lang
      newtype Title
                       = Title String
      newtype Author = Author String
      newtype Date
                       = Date String
      newtype Chapter = Chapter String
```

We have used the Haskell library HaXml [51], in particular the functionality in the module DtdToHaskell to obtain a data type from a DTD, together with functions for reading (parsing) and writing (pretty printing) valid XML documents to and from a value of the generated data type. For example, the following value of the above DTD:

```
<book lang="English">
<title> Dead Famous </title>
<author> Ben Elton </author>
<date> 2001 </date>
<chapter>Introduction </chapter>
<chapter>Preliminaries</chapter>
</book> ,
```

is translated to the following value of the data type Book:

```
Book\_Book\_Attrs \left\{ \begin{array}{l} bookLang = English \right\} \\ \left( Title ~ " \_ Dead \_ Famous \_ \square " \right) \\ \left( Author ~ " \_ Ben \_ Elton \_ \square \square " \right) \\ \left( Date ~ " \_ Duol \_ 2001 \_ \square \square \square \square " \right) \\ \left[ Chapter ~ "Introduction \_ " \\ , Chapter ~ "Preliminaries" \\ \right].
```

An element is translated to a value of a data type using just constructors and no labelled fields. An attribute is translated to a value that contains a labelled field for the attribute. Thus we can use the Generic Haskell constructs Con and Label to distinguish between elements and attributes in generic programs. We have not introduced the Label construct in these lecture notes. It it used to represent record labels in data types, and is very similar to the Con construct.

Separating structure and contents. The contents of an XML document is obtained by extracting all PCData (Parsable Character Data: characters without tags) and all CData (Character Data: characters with possibly tags, starting with '<! [CDATA[' ending in ']]>') from the document. In Generic Haskell, the contents of a value of a data type is obtained by extracting all strings from the value. For the above example value, we obtain the following result:

```
["_UDead_Famous_U", "_Ben_Elton_UUU", "_UBen_Elton_UUUU", "Introduction_U", "Preliminaries"].
```

The generic function *extract*, which extracts all strings from a value of a data type, is defined as follows:

```
 \begin{array}{lll} \textbf{type} \ \mathsf{Extract}\{\![\star]\!\} \ \mathsf{t} &= \mathsf{t} \to [\mathsf{String}] \\ \textbf{type} \ \mathsf{Extract}\{\![\kappa \to \nu]\!\} \ \mathsf{t} &= \forall \mathsf{a} \cdot \mathsf{Extract}\{\![\kappa]\!\} \ \mathsf{a} \to \mathsf{Extract}\{\![\nu]\!\} \ (\mathsf{t} \ \mathsf{a}) \\ extract\{\![\mathsf{Unit}]\!\} \ Unit &= [] \\ extract\{\![\mathsf{String}]\!\} \ s &= [s] \\ extract\{\![\mathsf{a} : + : \mathsf{b}]\!\} \ (Inl \ x) &= extract\{\![\mathsf{a}]\!\} \ x \\ extract\{\![\mathsf{a} : + : \mathsf{b}]\!\} \ (Inr \ y) &= extract\{\![\mathsf{b}]\!\} \ y \\ extract\{\![\mathsf{a} : * : \mathsf{b}]\!\} \ (x : * : y) &= extract\{\![\mathsf{a}]\!\} \ x + + extract\{\![\mathsf{b}]\!\} \ y \\ extract\{\![\mathsf{Con} \ c \ a]\!\} \ (Con \ b) &= extract\{\![\mathsf{a}]\!\} \ b. \end{array}
```

Note that it is possible to give special instances of a generic function on a particular type, as with $extract\{String\}$ in the above definition. Furthermore, because DtdToHaskell translates any DTD to a data type of kind \star , we could have defined extract just on data types of kind \star . However, higher-order kinds pose no problems. Finally, the operator + in the product case is a source of inefficiency. It can be removed using a standard transformation, see Exercise 8 in the first part of these lecture notes.

The structure from an XML document is obtained by removing all PCData and CData from the document. In Generic Haskell, the structure, or *shape*, of a value is obtained by replacing all strings by empty tuples. Thus we obtain a value that has a different type, in which occurrences of the type String have been replaced by the type (). This is another example of a type-indexed data type [23]. For example, the type we obtain from the data type Book is isomorphic to the

following data type:

```
\begin{array}{lll} {\rm data\ ShapeBook} &= ShapeBook\ ShapeBook\_Attrs \\ & ShapeAuthor \\ & ShapeAuthor \\ & ShapeDate \\ & [ShapeChapter] \\ {\rm data\ ShapeBook\_Attrs} &= ShapeBook\_Attrs \{ bookLang:: ShapeLang \} \\ {\rm data\ ShapeLang} &= SHAPEEnglish \mid SHAPEDutch \\ {\rm newtype\ ShapeTitle} &= ShapeTitle\ () \\ {\rm newtype\ ShapeAuthor} &= ShapeAuthor\ () \\ {\rm newtype\ ShapeChapter} &= ShapeChapter\ (), \end{array}
```

and the structure of the example value is

```
shapeBook = ShapeBook \ (ShapeBook\_Attrs\{bookLang = SHAPEEnglish\}) \\ (ShapeTitle\ ()) \\ (ShapeAuthor\ ()) \\ (ShapeDate\ ()) \\ [ShapeChapter\ () \\ , ShapeChapter\ () \\ ] \ .
```

The type-indexed data type Shape replaces occurrences of String in a data type by Unit.

```
\begin{split} &\mathsf{Shape}\{\mathsf{Unit}\} &= \mathsf{Unit} \\ &\mathsf{Shape}\{\mathsf{String}\} &= () \\ &\mathsf{Shape}\{a:+:b\} = \mathsf{Shape}\{a\}:+:\mathsf{Shape}\{b\} \\ &\mathsf{Shape}\{a:*:b\} = \mathsf{Shape}\{a\}:*:\mathsf{Shape}\{b\} \\ &\mathsf{Shape}\{\mathsf{Con}|a\} &= \mathsf{Con}(\mathsf{Shape}\{a\}) \end{split}
```

The generic function *shape* returns the shape of a value of any data type. It has the following kind-indexed type.

```
 \begin{array}{ll} \mathbf{type} \; \mathsf{Shape}\{\![\star]\!\} \; \mathsf{t} &= \mathsf{t} \to \mathsf{Shape}\{\![t]\!\} \\ \mathbf{type} \; \mathsf{Shape}\{\![\kappa \to \nu]\!\} \; \mathsf{t} = \forall \mathsf{a} \, . \, \mathsf{Shape}\{\![\kappa]\!\} \; \mathsf{a} \to \mathsf{Shape}\{\![\nu]\!\} \; (\mathsf{t} \; \mathsf{a}) \\ \end{array}
```

Note that we use the same name both for the kind-indexed type of function shape, as well as the type-indexed data type Shape . They can be distinguished based on their index.

Given the shape and the contents (obtained by means of function *extract*) of a value we obtain the original value by means of function *insert*:

```
insert\{|t:: \star|\}:: \mathsf{Shape}\{|t|\} \to [\mathsf{String}] \to \mathsf{t}.
```

The generic definition (with a kind-indexed type) of *insert* is left as an exercise.

Encoding constructors. The constructor of a value is encoded as follows. First calculate the number n of constructors of the data type. Then calculate the position of the constructor in the list of constructors of the data type. Finally, replace the constructor by the bit representation of its position, using $\log_2 n$ bits. For example, in a data type with 6 constructors, the third constructor is encoded by 010. We start counting with 0. Furthermore, a value of a data type with a single constructor is represented using 0 bits. Consequently, the values of all types except for String and Lang in the running example are represented using 0 bits.

We assume there exists a function *constructorPosition* which given a constructor returns a pair of integers: its position in the list of constructors of the data type, and the number of constructors of the data type.

```
constructorPosition :: ConDescr \rightarrow (Int, Int)
```

Function constructorPosition can be defined by means of function constructors, which returns the constructor descriptions of a data type. This function is defined in the module Collect, which can be found in the library of Generic Haskell.

```
constructors\{|t::\star|\}::[ConDescr]
```

Function *constructors* is defined for arbitrary kinds in module Collect. We omit the definitions of both function *constructors* and function *constructorPosition*.

The function *encode* takes a value, and encodes it as a value of type Bin, a list of bits, defined in the first part of these lecture notes. The difference with the function *encode* that is defined in the first part of these lecture notes is that here we encode the constructors of the value, and not the choices made in the sum. On average, the function *encode* given here compresses much better than the function *encode* from the first part of these lecture notes.

The interesting case in the definition of function *encode* is the constructor case. We first give the simple cases:

```
\begin{array}{lll} encode \{\!\!\mid t :: \kappa |\!\!\mid & :: \mathsf{Encode} \{\!\!\mid \kappa |\!\!\mid t \\ encode \{\!\!\mid \mathsf{Unit} \}\!\!\mid _- & = [] \\ encode \{\!\!\mid \mathsf{String} \}\!\!\mid _- & = [] \\ encode \{\!\!\mid \!\!a :*: b \}\!\!\mid (a :*: b) = encode \{\!\!\mid \!\!a \}\!\!\mid a + encode \{\!\!\mid \!\!b \}\!\!\mid b \\ encode \{\!\!\mid \!\!a :+: b \}\!\!\mid (Inl \ a) & = encode \{\!\!\mid \!\!a \}\!\!\mid a \\ encode \{\!\!\mid \!\!a :+: b \}\!\!\mid (Inr \ b) & = encode \{\!\!\mid \!\!b \}\!\!\mid b. \end{array}
```

For Unit and String there is nothing to encode. The product case encodes the components of the product, and concatenates the results. The sum case strips of the *Inl* or *Inr* constructor, and encodes the argument.

The encoding happens in the constructor case of function *encode*. We use function *intinrange2bits* to calculate the bits for the position of the argument constructor in the constructor list, given the number of constructors of the data type currently in scope. The definition of *intinrange2bits* is omitted.

```
encode\{\{Con\ c\ a\}\}\ (Con\ a) = intinrange2bits\ (constructorPosition\ c) + encode\{\{a\}\}\ a intinrange2bits::(Int,Int) \rightarrow Bin
```

We omit the definition of the function to decode a list of bits into a value of a data type. This function is the inverse of function *encode* defined in this section, and is very similar to the function *decodes* given in the first part of these lecture notes.

Compressing the contents. Finally, it remains to compress the contents of an XML document. At the moment we use the Unix compress utility [52] to compress the strings obtained from the document.

3.2 Analysis

How does XCOMPREZ perform, and how does it compare with other XML compressors? The analysis given in this section is limited: XCOMPREZ is used as an example of a generic program, and not as the latest development in XML compression. Furthermore, we have not been able to obtain the executables or the source code of most of the existing XML compressors.

Existing XML compressors. Structure-specific compression methods give much better compression results [4, 15, 14, 46] than conventional compression methods such as the Unix compress utility [52]. There exist many XML compressors; we know of XMLZip [12], XMill [36], ICT's XML-Xpress [25], Millau [16], XMLPPM [6], XGrind [48], and lossy XML compression [5]. We will not perform an exhaustive comparison between our compressor and these compressors, but we will briefly compare our compressor with XMill.

Compression ratio. We have performed some initial tests comparing XCOMPREZ and XMill. The tests are not representative, and it is impossible to draw hard conclusions from the results. However, on our test examples XCOMPREZ is 40% to 50% better than XMill. We think this improvement in compression ratio is considerable. When we replace HaXml by a tool that generates a data type for a schema, we expect that we can achieve better compression ratios.

Code size. With respect to code size, the difference between XMill and XCOM-PREZ is dramatic: XMill is written in almost 20k lines of C++. The main functionality of XCOMPREZ is less than 300 lines of Generic Haskell code. Of course, for a fair comparison we have to add some of the HaXml code (which is a library distributed together with almost all compiler and interpreters for Haskell), the code for handling bits, and the code for implementing the as yet unimplemented features of XMill. We expect to be able implement all of XMill's features in about 20% of the code size of XMill.

Extensions of XCOMPREZ. A relatively simple way to improve XCOMPREZ it is to analyze some source files that are valid with respect to the DTD, count the number of occurrences of the different elements (constructors), and apply Huffman coding. We have implemented this rather simple extension [32].

XMill stores the strings in the elements in so-called containers. The standard approach in XMill is to use different containers for different elements, so that, for example, all authors are stored in the author container, all dates in the date container, etc. Since the strings in for example the container for dates are very similar, standard compression methods can compress the containers with a larger factor than the single file obtained by storing all strings that appear in the XML document. Again, it is easy to implement this feature as a generic program [32].

Finally, we have also used (Adaptive) Arithmetic Coding [3] to compress the constructors.

Specializing XCOMPREZ. XCOMPREZ can not only be used for compressing XML documents, but also for compressing values of arbitrary data types that have not necessarily been generated by DtdToHaskell.

Suppose we have a data type that contains a constructor Age, which takes an integer as argument and denotes the age of a human being. Since 128 is currently a safe upperbound for the age of a human being, it suffices to use 7 bits for the integer argument of Age. Suppose compressAge calculates these 7 bits from an age. Then we can reuse the definition of encode together with a constructor case [10] to define a function specialEncode.

```
\begin{array}{lll} specialEncode\{\!\!\mid\!\! t:: \kappa\} & :: \mathsf{Encode}\{\!\!\mid\!\! \kappa\} \mathsf{\,t} \\ specialEncode\{\!\!\mid\!\! \mathbf{case} \; Age\} \mathsf{\,(Age} \; i) = compressAge \; i \\ specialEncode\{\!\!\mid\!\! \mathbf{a}\} & = encode\{\!\!\mid\!\! \mathbf{a}\} \end{array}
```

Function specialEncode is still a generic encoding function, but on values of the form Age i it uses a different compress function.

3.3 Conclusions

We have shown how to implement an XML compressor as a generic program. XCompresses better than for example XMill because it uses the information about an XML document present in a DTD.

There exist several other classes of XML tools that can be implemented as generic programs, and that would benefit from such an implementation. Examples of such tools are XML editors and XML databases [17]. The combination of HaXml and generic programming as in Generic Haskell is very useful for implementing the kind of XML tools for which DTDs play an important rôle. Using generic programming, such tools become easier to write, because a lot of the code pertaining to DTD handling and optimisation is obtained from the generic programming compiler, and the resulting tools are more effective, because they directly depend on the DTD. For example, a DTD-aware XML compressor, such as XCOMPREZ described in this paper, compresses considerably better than XML compressors that don't take the DTD into account, such as XMill. Furthermore, our compressor is much smaller than XMill.

Although we think Generic Haskell is very useful for developing DTD-aware XML tools, there are some features of XML tools that are difficult to express in Generic Haskell. Some of the functionality in the DOM, such as the methods childNodes and firstChild in the Node interface, is hard to express in a typed way. A flexible extension of type-indexed data types [23] might offer a solution to this problem. We believe that fusing HaXml, or a tool based on Schemas, with Generic Haskell, obtaining a 'domain-specific' language [8] for generic programming on DTDs or Schemas is a promising approach.

For tools that do not depend on a DTD we can use the untyped approach from HaXml to obtain a tool that works for any document. However, most of the advantages of generic programming no longer apply.

Exercise 3. Adapt the function extract such that it returns a list of containers, where a container is a list of strings. Return a (possibly empty) container for every constructor name.

Exercise 4. There might be many empty containers when using the approach from the previous exercise. Analyse a data type for occurrences of the type String under constructors. Use this analysis to only return containers for constructor names that might contain strings.

Exercise 5. Function *insert* takes the shape and the contents (a list of strings) of a value, and inserts the strings at the right positions in the shape. Define function *insert* as a a generic function with a kind-indexed type.

Exercise 6. Adapt the current version of XCOMPREZ such that it can use Huffman coding instead of the standard constructor encoding used in this section. Make sure other encodings can be used as well.

4 The zipper

This section shows how to define a so-called zipper for an arbitrary data type. This is an advanced example demonstrating the full power of a type-indexed data type together with a number of generic functions working on it.

The zipper is a data structure that is used to represent a tree together with a subtree that is the focus of attention, where that focus may move left, right, up or down in the tree. The zipper is used in tools where a user interactively manipulates trees, for instance, in editors for structured documents such as proofs or programs. For the following it is important to note that the focus of the zipper may only move to recursive components. Consider as an example the data type Tree:

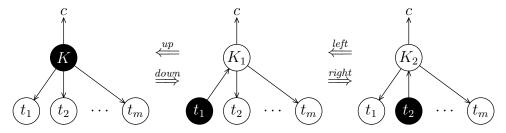
$$\mathbf{data}$$
 Tree $\mathsf{a} \mathsf{b} = Tip \mathsf{a} \mid Node \text{ (Tree a b) b (Tree a b)}.$

If the left subtree of a *Node* constructor is the current focus, moving right means moving to the right tree, not to the b-label. This implies that recursive positions in trees play an important rôle in the definition of a generic zipper data structure. To obtain access to these recursive positions, we have to be explicit about the fixed points in data type definitions. The zipper data structure is then defined by induction on the so-called pattern functor of a data type.

The tools in which the zipper is used, allow the user to repeatedly apply navigation or edit commands, and to update the focus accordingly. In this section we define a type-indexed data type for locations, which consist of a subtree (the focus) together with a context, and we define several navigation functions on locations.

4.1 The basic idea

The zipper is based on pointer reversal. If we follow a pointer to a subterm, the pointer is reversed to point from the subterm to its parent so that we can go up again later. A location is a pair (t,c) consisting of the current subterm t and a pointer c to its parent. The upward pointer corresponds to the context of the subterm. It can be represented as follows. For each constructor K that has m recursive subcomponents we introduce m context constructors K_1, \ldots, K_m . Now, consider the location $(K \ t_1 \ t_2 \ \ldots \ t_m, c)$. If we go down to t_1 , we are left with the context $K \bullet t_2 \ \ldots \ t_m$ and the old context c. To represent the combined context, we simply plug c into the hole to obtain $K_1 \ c \ t_2 \ \ldots \ t_m$. Thus, the new location is $(t_1, K_1 \ c \ t_2 \ \ldots \ t_m)$. The following picture illustrates the idea (the filled circle marks the current cursor position).



4.2 Data types as fixed points of pattern functors

As mentioned above, in order to use the zipper, we have to be explicit about the fixed points in data type definitions. Therefore, we introduce the data type Fix,

which is used to define a data type as a fixed point of a pattern functor. The pattern functor makes the recursion explicit in a data type.

```
newtype Fix f = In\{out :: f(Fix f)\}
```

This is a labelled variant of the data type Fix defined in Section 1.2 of the first part of these lecture notes. For example, the data types of natural numbers and bushes can be defined using explicit fixed points as follows:

```
 \begin{array}{ll} \mathbf{data} \ \mathsf{NatF} \ \mathsf{a} &= ZeroF \mid SuccF \ \mathsf{a} \\ \mathbf{type} \ \mathsf{Nat} &= \mathsf{Fix} \ \mathsf{NatF} \\ \mathbf{data} \ \mathsf{BushF} \ \mathsf{a} &= LeafF \ \mathsf{Char} \mid BinF \ \mathsf{a} \ \mathsf{a} \\ \mathbf{type} \ \mathsf{Bush} &= \mathsf{Fix} \ \mathsf{BushF}. \end{array}
```

It is easy to convert between data types defined as fixed points and the original data type definitions of natural numbers and bushes. However, nested data types and mutually recursive data types cannot be defined in terms of this particular definition of Fix.

4.3 Type indices of higher kinds

The types that occur in the indices of a generic function have kind \star as their base kind. For example, Int, Char and Unit are all of kind \star , and :+: and :*: have kind $\star \to \star \to \star$. In this section we are going to define generic functions which have $\star \to \star$ as their base kind. We need slightly different type indices for generic functions operating on types of kind $\star \to \star$:

```
\begin{array}{lll} {\sf K} \; {\sf t} &= \varLambda {\sf a} \, . \, {\sf t} \\ {\sf f1} : + : \; {\sf f2} = \varLambda {\sf a} \, . \, {\sf f1} \; {\sf a} : + : \; {\sf f2} \; {\sf a} \\ {\sf f1} : * : \; {\sf f2} = \varLambda {\sf a} \, . \, {\sf f1} \; {\sf a} : * : \; {\sf f2} \; {\sf a} \\ {\sf Con} \; c \; {\sf f} &= \varLambda {\sf a} \, . \, {\sf Con} \; c \; ({\sf f} \; {\sf a}) \\ {\sf Id} &= \varLambda {\sf a} \, . \; {\sf a}. \end{array}
```

We have the constant functor K, which lifts a type of kind \star to kind $\star \to \star$. We will need K Unit as well as K Char (or more general, K t for all primitive types). We overload :+:, :*:, and Con, to be lifted versions of their previously defined counterparts. The only new type index in this set of indices of kind $\star \to \star$ is the identity functor Id. Hinze [20] shows that these types are the normal forms of types of kind $\star \to \star$.

4.4 Locations

A location is a subtree, together with a context, which encodes the path from the top of the original tree to the selected subtree. The type-indexed data type Loc returns a type for locations given an argument pattern functor.

```
\begin{array}{lll} \mathsf{Loc}\{\!\!\lceil f :: \star \to \star \}\!\!\} & :: & \star \\ \mathsf{Loc}\{\!\!\lceil f \}\!\!\} & = & (\mathsf{Fix}\,\mathsf{f}, \mathsf{Context}\{\!\!\lceil f \}\!\!\} \; (\mathsf{Fix}\,\mathsf{f})) \\ \mathsf{Context}\{\!\!\lceil f :: \star \to \star \}\!\!\} & :: & \star \to \star \\ \mathsf{Context}\{\!\!\lceil f \}\!\!\} \; r & = & \mathsf{Fix}\; (\mathsf{LMaybe}\; (\mathsf{Ctx}\{\!\!\lceil f \}\!\!\} \; r)) \\ \mathsf{data}\; \mathsf{LMaybe}\; \mathsf{f}\; \mathsf{a} & = & \mathit{LNothing}\; |\; \mathit{LJust}\; (\mathsf{f}\; \mathsf{a}), \end{array}
```

where LMaybe is the lifted version of Maybe. The type Loc is defined in terms of Context, which constructs the context parameterized by the original tree type. The Context of a value is either empty (represented by *LNothing* in the LMaybe type), or it is a path from the root down into the tree. Such a path is constructed by means of the argument type of LMaybe: the type-indexed data type Ctx. The type-indexed data type Ctx is defined by induction on the pattern functor of the original data type. It can be seen as the *derivative* (as in calculus) of the pattern functor f [39, 1]. If the derivative of f is denoted by f', we have

$$\begin{aligned} &const' &= 0 \\ &(f+g)' &= f'+g' \\ &(f\times g)' &= f'\times g+f\times g'. \end{aligned}$$

It follows that in the definition of Ctx we will also need access to the type arguments themselves on the right-hand side of the definition.

This definition can be understood as follows. Since it is not possible to descend into a constant, the constant cases do not contribute to the result type, which is denoted by the 'empty type' Void, a type without values. The ld case denotes a recursive component, in which it is possible to descend. Hence it may occur in a context. Descending in a value of a sum type follows the structure of the input value. Finally, there are two ways to descend in a product: descending left, adding the contents to the right of the node to the context, or descending right, adding the contents to the left of the node to the context.

For example, for natural numbers with pattern functor K Unit:+: Id, and for trees of type Bush with pattern functor BushF, which can be represented by K Char:+:(Id:*: Id) we obtain

The context of a natural number is isomorphic to a natural number (the context of m in n is n-m), and the context of a Bush applied to the data type Bush itself is isomorphic to the type Context_Bush introduced in Section 1.

McBride [39,1] also defines a type-indexed zipper data type. His zipper slightly deviates from Huet's and our zipper: the navigation functions on McBride's zipper are not constant time anymore. The observation that the Context of a data type is its derivative (as in calculus) is due to McBride.

4.5 Navigation functions

We define generic functions on the type-indexed data types Loc, Context, and Ctx for navigating through a tree. All of these functions act on locations. These are the basic functions for the zipper.

Function down. The function down is a generic function that moves down to the leftmost recursive child of the current node, if such a child exists. Otherwise, if the current node is a leaf node, then down returns the location unchanged.

$$down\{f :: \star \to \star\} :: \mathsf{Loc}\{f\} \to \mathsf{Loc}\{f\}$$

The instantiation of down to the data type Bush has been given in Section 1. The function down satisfies the following property:

$$\forall m . down\{\{f\}\} \ m \neq m \implies (up\{\{f\}\} \cdot down\{\{f\}\}) \ m = m,$$

where the function up goes up in a tree. So first going down the tree and then up again is the identity function on locations in which it is possible to go down.

Since down moves down to the leftmost recursive child of the current node, the inverse equality $down\{f\} \cdot up\{f\} = id$ does not hold in general. However, there does exist a natural number n such that

$$\forall m . up\{f\} \ m \neq m \implies (right\{f\}^n \cdot down\{f\} \cdot up\{f\}) \ m = m,$$

where the function right goes right in a tree. These properties do not completely specify function down. The other properties it should satisfy are that the selected subtree of $down\{f\}$ m is the leftmost tree-child of the selected subtree of m, and the context of $down\{f\}$ m is the context of m extended with all but the leftmost tree-child of m.

The function down is defined as follows.

$$down\{\{f\}\}\ (t,c) = \mathbf{case}\ first\{\{f\}\}\ (out\ t)\ c\ \mathbf{of}$$

$$Just\ (t',c') \to (t',In\ (LJust\ c'))$$

$$Nothing \to (t,c)$$

To find the leftmost recursive child, we have to pattern match on the pattern functor f, and find the first occurrence of ld. The helper function first is a generic function that possibly returns the leftmost recursive child of a node, together with the context (a value of type $Ctx\{f\}\ c\ t$) of the selected child. The function

down then turns this context into a value of type Context by inserting it in the right ('non-top') component of a sum by means of LJust, and applying the fixed point constructor In to it.

```
\begin{array}{lll} \mathit{first} \{ \mathsf{f} :: \star \to \star \} & :: \ \forall \mathsf{c} \ \mathsf{t} . \mathsf{f} \ \mathsf{t} \to \mathsf{c} \to \mathsf{Maybe} \ (\mathsf{t}, \mathsf{Ctx} \{ \mathsf{f} \} \ \mathsf{c} \ \mathsf{t}) \\ \mathit{first} \{ \mathsf{Id} \} \ \mathit{t} \ \mathit{c} & = \mathit{return} \ (\mathit{t}, \mathit{c}) \\ \mathit{first} \{ \mathsf{K} \ \mathsf{Unit} \} \ \mathit{t} \ \mathit{c} & = \mathit{Nothing} \\ \mathit{first} \{ \mathsf{K} \ \mathsf{Char} \} \ \mathit{t} \ \mathit{c} & = \mathit{Nothing} \\ \mathit{first} \{ \mathsf{f1} :+: \ \mathsf{f2} \} \ (\mathit{Inl} \ \mathit{x}) \ \mathit{c} & = \ \mathsf{do} \ \{ (\mathit{t}, \mathit{cx}) \leftarrow \mathit{first} \{ \mathsf{f1} \} \ \mathit{x} \ \mathit{c}; \mathit{return} \ (\mathit{t}, \mathit{Inl} \ \mathit{cx}) \} \\ \mathit{first} \{ \mathsf{f1} :+: \ \mathsf{f2} \} \ (\mathit{Inr} \ \mathit{y}) \ \mathit{c} & = \ \mathsf{do} \ \{ (\mathit{t}, \mathit{cy}) \leftarrow \mathit{first} \{ \mathsf{f2} \} \ \mathit{y} \ \mathit{c}; \mathit{return} \ (\mathit{t}, \mathit{Inr} \ \mathit{cy}) \} \\ \mathit{first} \{ \mathsf{f1} :+: \ \mathsf{f2} \} \ (\mathit{x} :+: \ \mathit{y}) \ \mathit{c} & = \ \mathsf{do} \ \{ (\mathit{t}, \mathit{cy}) \leftarrow \mathit{first} \{ \mathsf{f2} \} \ \mathit{y} \ \mathit{c}; \mathit{return} \ (\mathit{t}, \mathit{Inl} \ (\mathit{cx}, \mathit{y})) \} \\ \\ + \ \mathsf{do} \ \{ (\mathit{t}, \mathit{cy}) \leftarrow \mathit{first} \{ \mathsf{f2} \} \ \mathit{y} \ \mathit{c}; \mathit{return} \ (\mathit{t}, \mathit{Inr} \ (\mathit{x}, \mathit{cy})) \} \\ \end{array}
```

Here, return is obtained from the Maybe monad, and the operator (#) is the standard monadic plus, called mplus in Haskell, given by

```
\begin{array}{ll} (++) & :: \ \forall {\it a} \ . \ {\it Maybe} \ {\it a} \to {\it Maybe} \ {\it a} \\ Nothing ++ m &= m \\ Just \ a ++ m &= Just \ a. \end{array}
```

The function *first* returns the value and the context at the leftmost ld position. So in the product case, it first tries the left component, and only if it fails, it tries the right component.

The definitions of functions up, right and left are not as simple as the definition of down, since they are defined by pattern matching on the context instead of on the tree itself. We will just define functions up and right, and leave function left as an exercise.

Function up. The function up moves up to the parent of the current node, if the current node is not the top node.

```
\begin{array}{ll} up\{\!\mid\!\!f\colon \star\to\star\}\ ::\ \mathsf{Loc}\{\!\mid\!\!f\}\to\mathsf{Loc}\{\!\mid\!\!f\}\}\\ up\{\!\mid\!\!f\}\ (t,c) &=\ \mathsf{case}\ out\ c\ \mathsf{of}\\ & LNothing\to(t,c)\\ & LJust\ c'\to\mathsf{do}\ \{\mathit{ft}\leftarrow\mathit{insert}\{\!\mid\!\!f\}\}\ c'\ t;\\ & c''\leftarrow\mathit{extract}\{\!\mid\!\!f\}\!\!\mid\! c';\\ & \mathit{return}\ (\mathit{In}\ \mathit{ft},c'')\} \end{array}
```

Remember that LNothing denotes the empty top context. The navigation function up uses two helper functions: insert and extract. The latter returns the context of the parent of the current node. Each element of type $Ctx\{f\}\$ c t has at most one c component (by an easy inductive argument), which marks the context of the parent of the current node. The generic function extract extracts

this context.

```
\begin{array}{lll} \textit{extract}\{ | \mathsf{f} :: \star \to \star \} & :: \; \forall \mathsf{c} \; \mathsf{t} \cdot \mathsf{Ctx}\{ | \mathsf{f} \} \; \mathsf{c} \; \mathsf{t} \to \mathsf{Maybe} \; \mathsf{c} \\ \textit{extract}\{ | \mathsf{Id} \} \; c & = \textit{return} \; c \\ \textit{extract}\{ | \mathsf{K} \; \mathsf{Unit} \} \; c & = \textit{Nothing} \\ \textit{extract}\{ | \mathsf{f} \; \mathsf{ct} | \mathsf{ct}
```

The function *extract* is polymorphic in c and in t.

Function *insert* takes a context and a tree, and inserts the tree in the current focus of the context, effectively turning a context into a tree.

```
 \begin{array}{lll} insert\{ | \mathsf{f} :: \star \to \star \} & :: \; \forall \mathsf{c} \; \mathsf{t} . \; \mathsf{Ctx}\{ | \mathsf{f} \} \; \mathsf{c} \; \mathsf{t} \to \mathsf{t} \to \mathsf{Maybe} \; (\mathsf{f} \; \mathsf{t}) \\ insert\{ | \mathsf{Id} \} \; c \; t & = return \; t \\ insert\{ | \mathsf{K} \; \mathsf{Unit} \} \; c \; t & = Nothing \\ insert\{ | \mathsf{K} \; \mathsf{Char} \} \; c \; t & = Nothing \\ insert\{ | \mathsf{f} \; 1 :+ : \mathsf{f} \; \mathsf{2} \} \; (Inl \; cx) \; t & = \mathbf{do} \; \{ x \leftarrow insert\{ | \mathsf{f} \; \mathsf{1} \} \; cx \; t; return \; (Inl \; x) \} \\ insert\{ | \mathsf{f} \; 1 :+ : \mathsf{f} \; \mathsf{2} \} \; (Inl \; (cx, y)) \; t & = \mathbf{do} \; \{ x \leftarrow insert\{ | \mathsf{f} \; \mathsf{1} \} \; cx \; t; return \; (x, y) \} \\ insert\{ | \mathsf{f} \; 1 :+ : \mathsf{f} \; \mathsf{2} \} \; (Inl \; (cx, y)) \; t & = \mathbf{do} \; \{ y \leftarrow insert\{ | \mathsf{f} \; \mathsf{2} \} \; cy \; t; return \; (x, y) \} \\ insert\{ | \mathsf{f} \; 1 :+ : \mathsf{f} \; \mathsf{2} \} \; (Inr \; (x, cy)) \; t & = \mathbf{do} \; \{ y \leftarrow insert\{ | \mathsf{f} \; \mathsf{2} \} \; cy \; t; return \; (x, y) \} \\ \end{array}
```

Note that the extraction and insertion is happening in the identity case Id; the other cases only pass on the results.

Since $up\{f\} \cdot down\{f\} = id$ on locations in which it is possible to go down, we expect similar equalities for the functions first, extract, and insert. We have that the following computation

```
\begin{aligned} \mathbf{do} & \{(t,c') \leftarrow \mathit{first} \{ \mathsf{f} \} \ \mathit{ft} \ c; \\ & c'' \leftarrow \mathit{extract} \{ \mathsf{f} \} \ c'; \\ & \mathit{ft'} \leftarrow \mathit{insert} \{ \mathsf{f} \} \ c' \ t \\ & \mathit{return} \ (c == c'' \land \mathit{ft} == \mathit{ft'} \ ) \ \}, \end{aligned}
```

returns true on locations in which it is possible to go down.

Function right. The function right moves the focus to the next (right) sibling in a tree, if it exists. The context is moved accordingly. The instance of right on the data type Bush has been given in Section 1. The function right satisfies the following property:

```
\forall m . right\{|f|\} \ m \neq m \implies (left\{|f|\} \cdot right\{|f|\}) \ m = m,
```

that is, first going right in the tree and then left again is the identity function on locations in which it is possible to go to the right. Of course, the dual equality holds on locations in which it is possible to go to the left. Furthermore, the selected subtree of $right\{f\}$ m is the sibling to the right of the selected subtree

of m, and the context of $right\{f\}$ m is the context of m in which the context is replaced by the selected subtree of m, and the first subtree to the right of the context of m is replaced by the context of m.

Function *right* is defined by pattern matching on the context. It is impossible to go to the right at the top of a tree. Otherwise, we try to find the right sibling of the current focus.

```
\begin{array}{ll} right \{\!\!\lceil f :: \star \to \star \}\!\!\mid :: & \mathsf{Loc} \{\!\!\lceil f \}\!\!\mid \to \mathsf{Loc} \{\!\!\lceil f \}\!\!\mid \\ right \{\!\!\lceil f \}\!\!\mid (t,c) &= & \mathsf{case} \ out \ c \ \mathsf{of} \\ & & LNothing \to (t,c) \\ & & LJust \ c' \to \mathsf{case} \ next \{\!\!\lceil f \}\!\!\mid t \ c' \ \mathsf{of} \\ & & Just \ (t',c'') \to (t',In \ (LJust \ c'')) \\ & & & Nothing \to (t,c) \end{array}
```

The helper function next is a generic function that returns the first location that has the recursive value to the right of the selected value as its focus. Just as there exists a function left such that $left\{f\} \cdot right\{f\} = id$ (on locations in which it is possible to go to the right), there exists a function previous, such that

```
 \begin{aligned} \mathbf{do} \; \{(t',c') \leftarrow next \| \mathbf{f} \| \; t \; c \; ; \\ (t'',c'') \leftarrow previous \| \mathbf{f} \| \; t' \; c'; \\ return \; (c = c'' \wedge t = t'') \}, \end{aligned}
```

returns *true* (on locations in which it is possible to go to the right). We will define function *next*, and omit the definition of function *previous*.

```
next\{f:: \star \to \star\}:: \forall c \ t \cdot t \to Ctx\{f\} \ c \ t \to Maybe \ (t, Ctx\{f\} \ c \ t)
                                             = Nothing
next\{|\mathsf{Id}|\} \ t \ c
next\{|K|Unit\}\} t c
                                             = Nothing
next\{|K|Char|\} t c
                                             = Nothing
next\{f1:+: f2\}\ t\ (Inl\ cx)
   = \mathbf{do} \{ (t', cx') \leftarrow next \{ | \mathbf{f1} | t \ cx; return \ (t', Inl \ cx') \} 
next\{f1:+: f2\}\ t\ (Inr\ cy)
    = \mathbf{do} \{ (t', cy') \leftarrow next \{ | f2 \} \ t \ cy; return \ (t', Inr \ cy') \}
next\{|f1:*:f2|\}\ t\ (Inl\ (cx,y)\ )
   = \mathbf{do} \{ (t', cx') \leftarrow next \{ f1 \} \ t \ cx; return \ (t', Inl \ (cx', y)) \}
                ++do { c \leftarrow extract\{|f1|\} \ cx;
                           x \leftarrow insert\{\{f1\}\}\ cx\ t;
                           (t', cy) \leftarrow first\{|f2|\} \ y \ c;
                           return (t', Inr (x, cy))
next\{ | f1 : * : f2 \} \ t \ (Inr \ (x, cy))
   = \mathbf{do} \left\{ (t', cy') \leftarrow next \{ \{f2\} \ t \ cy; return \ (t', Inr \ (x, cy')) \} \right\}
```

The first three lines in this definition show that it is impossible to go to the right in an identity or constant context. If the context argument is a value of a sum, we select the next element in the appropriate component of the sum. The product case is the most interesting one. If the context is in the right component of a pair, next returns the next value of that context, properly combined with

the left component of the tuple. On the other hand, if the context is in the left component of a pair, the next value may be either in that left component (the context), or it may be in the right component (the value). If the next value is in the left component, it is returned by the first line in the definition of the product case. If it is not, next extracts the context c (the context of the parent) from the left context cx, it inserts the given value in the context cx giving a 'tree' value x, and selects the first component in the right component of the pair, using the extracted context c for the new context. The new context that is thus obtained is combined with x into a context for the selected tree.

Exercise 7. Define the function left:

$$left\{|f::\star\to\star|\}::Loc\{|f|\}\to Loc\{|f|\}.$$

Exercise 8. If you don't want to use the zipper, you can alternatively keep track of the path to the current focus. Suppose we want to use the path to determine the name of the top constructor of the current focus in a value of a data type. The path determines which child of a value is selected. Since the products used in our representations of data types are binary, a path has the following structure:

data
$$Dir = L \mid R$$

type $Path = [Dir]$.

The to be defined function *selectCon* takes a value of a data type and a path, and returns the constructor name at the position denoted by the path. For example,

```
 \begin{split} & \mathbf{data} \ \mathsf{List} \ = \mathit{Nil} \ | \ \mathit{Cons} \ \mathsf{Char} \ \mathsf{List} \\ & \mathit{selectCon} \{ \mathsf{List} \} \ (\mathit{Cons} \ 2 \ (\mathit{Cons} \ 3 \ (\mathit{Cons} \ 6 \ \mathit{Nil}))) \ [\mathit{R}, \mathit{R}, \mathit{R}] \\ & \Longrightarrow "\mathtt{Nil}" \\ & \mathbf{data} \ \mathsf{Tree} = \mathit{Leaf} \ \mathsf{Int} \ | \ \mathit{Node} \ \mathsf{Tree} \ \mathsf{Int} \ | \ \mathit{Node} \ (\mathit{Leaf} \ 1) \ 3 \ (\mathit{Node} \ (\mathit{Leaf} \ 2) \ 1 \ (\mathit{Leaf} \ 5))) \ [\mathit{R}, \mathit{R}] \\ & \Longrightarrow "\mathtt{Node}" \\ & \mathit{selectCon} \{ \mathsf{Tree} \} \ (\mathit{Node} \ (\mathit{Leaf} \ 1) \ 3 \ (\mathit{Node} \ (\mathit{Leaf} \ 2) \ 1 \ (\mathit{Leaf} \ 5))) \ [\mathit{R}, \mathit{R}, \mathit{L}] \\ & \Longrightarrow "\mathtt{Leaf}". \end{split}
```

Define the generic function selectCon, together with its kind-indexed type.

Exercise 9. Define the function *left*, which takes a location, and returns the location to the left of the argument location, if possible.

Exercise 10. For several applications we have to extend a data type such that it is possible to represent a place holder. For example, from the data type Tree defined by

data Tree a b =
$$Tip$$
 a | $Node$ (Tree a b) b (Tree a b),

we would like to obtain a type isomorphic to the following type:

```
data HoleTree a b = Hole \mid Tip a | Node (HoleTree a b) b (HoleTree a b).
```

- Define a type-indexed data type Hole that takes a data type and returns a data type in which also holes can be specified. Also give the kind-indexed kind of this type-indexed data type. (The kind-indexed kind cannot and does not have to be defined in Generic Haskell though.)
- Define a generic function to Hole which translates a value of a data type t to a value of the data type Hole \{t\}, and a function from Hole that does the inverse for values that do not contain holes anymore:

```
 \begin{array}{lll} toHole\{\![t::\kappa]\!\} & :: & \mathsf{ToHole}\{\![\kappa]\!\} \ t \\ fromHole\{\![t::\kappa]\!\} & :: & \mathsf{FromHole}\{\![\kappa]\!\} \ t \\ \\ \mathbf{type} \ \mathsf{ToHole}\{\![\star]\!\} \ t & = t \to \mathsf{Hole}\{\![t]\!\} \\ \mathbf{type} \ \mathsf{FromHole}\{\![\star]\!\} \ t & = \mathsf{Hole}\{\![t]\!\} \to t. \end{array}
```

5 Conclusions

We have developed three advanced applications in Generic Haskell. In these examples we use, besides generic functions with kind-indexed kinds, type-indexed data types, dependencies between and generic abstractions of generic functions, and default and constructor cases. Some of the latest developments of Generic Haskell have been guided by requirements from these applications.

We hope to develop more applications using Generic Haskell in the future, both to develop the theory and the language. Current candidate applications are more XML tools and editors.

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