# Investigating Minimally Strict Functions in Functional Programming 

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## 1 Introduction

In this thesis we consider the non-strict, functional programming language Haskell (Peyton Jones 2003). First of all, functional programing languages belong to the class of declarative programming languages. While in an imperative programming language the programmer specifies a procedure to get a solution to a problem, in a declarative language the programmer instead specifies the structure of a solution. In other words, while in an imperative language we define how a solution is computed, in a declarative language we rather define what is computed. Besides, Haskell has a quite strong type system, which is able to catch some bugs at compile-time, while these bugs do not emerge until run-time in a language with a weaker type system or no type system at all. And, finally, features like higher-order functions and overloading allow programmers to reuse code, which results in programs that are smaller and easier to maintain.

However, even with a comparatively high-level language like Haskell, programmers may write programs that are not optimal with respect to a certain criterion. Obviously, the most important criterion is correctness. That is, we prefer a program that meets our specifications over a program that does not. There are several approaches that assist programmers in writing correct Haskell programs like prop-erty-based testing (Claessen and Hughes 2000) or even automatic proving (Sonnex et al. 2011). Nonetheless, there are other important criteria to rate programs. For example, we could classify several implementations of the same algorithm in one language by means of run-time or memory consumption. This thesis investigates another criterion, namely, rating Haskell programs by means of strictness.

Intuitively, in contrast to a strict programming language, in a non-strict programming language like Haskell a function only inspects the part of its argument that is needed to yield a certain part of its result. In the extreme case of a constant function, for example, the function does not evaluate its argument at all as the result of the function does not depend on the argument. Nevertheless, we can still define a constant function in Haskell that unnecessarily inspects its argument. More generally speaking, in Haskell a function definition rather non-declaratively specifies how an argument is inspected. There are often several implementations of functions with the same basic behavior that inspect different parts of their argument. While in the case of a constant function it is quite easy to observe that it does not have to inspect its argument, we will see that in general it is more difficult to observe whether a function can yield the same result by inspecting a smaller part of its argument.

To rate functions with respect to strictness, we consider the following definition. A function $f$ is less strict than a function $g$ if the function $f$ never inspects a larger part of its argument than the function $g$ does, to yield a certain result. In addition, there exists a result such that $f$ inspects a strictly smaller part of its argument than $g$ does. We call a function $f$ with these properties less strict than $g$. This "definition" is quite informal, and we will later give a formal definition of the less-strict relation by

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means of a denotational semantics. For now this intuitive view should be sufficient for this introduction and will be helpful later on.

As an example for the interest in less strict functions, let us consider the standard Haskell function partition. Waldmann (2000) as well as Russell (2000) have stated that the implementation of this function was more strict than necessary. The function partition takes a predicate - a unary funtion that yields a Boolean value - and a list - an ordered collection of values of equal type - and yields a pair - a container for two values. The pair holds two lists where the first list contains the elements that satisfy the predicate and the second list contains the elements that do not. It has been observed that partition, at that time, was unnecessarily strict because the implementation did not comply with an abstract specification given in the Haskell language report (Peyton Jones 2003).

The language report specifies the behavior of partition by means of the standard Haskell function filter. The function filter takes a list and a predicate and yields a list of all elements that satisfy the predicate. The behavior of partition is specified by the following equivalence where $p$ is an arbitrary predicate and $x s$ an arbitrary list. Furthermore, the symbol $\equiv$ denotes that the left-hand and the right-hand side yield the same results when evaluated.

$$
\text { partition } p x s \equiv(\text { filter } p x s, \text { filter }(\text { not } \circ p) x s)
$$

The application of a function is denoted by juxta-position, for example, partition $p x s$ denotes the application of the function partition to the arguments $p$ and $x s$. The parenthesis together with the comma on the right-hand side of the equivalence denote the construction of a pair. That is, partition applied to some predicate $p$ and a list $x$ s yields the same result as the pair that consists of an application of filter to the predicate and the list and an application of filter to the negation of the predicate and the list. The operator o denotes function composition and not is the Boolean negation, so the term not op denotes a predicate that first applies the original predicate $p$ and negates its Boolean result.

Waldmann (2000) named the following example, which demonstrates, that the original implementation of partition does not comply with the behavior of two applications of filter.

$$
\text { let }(x s, y s)=\text { partition odd }[1 . .] \text { in (take } 3 x s \text {, take } 3 y s)
$$

The expression [1..] denotes an infinite list of increasing natural numbers that starts with one. Because Haskell is a non-strict programming language we can use infinite data structures as a function might only inspect a small part of the infinite structure. The infinite list is partitioned into odd and even numbers by using partition and the predicate odd, which is satisfied if its argument is an odd number. The let expression simply introduces names for the components of the result, which is a pair, of the function partition. By equating a pair constructed of the variables $x s$ and $y s$ with the application partition odd [1..], the variables $x s$ and $y s$ become names for the first and the second component of the result of partition, respectively. We can use the variables $x s$ and $y s$ on the right-hand side of the keyword in as abbreviations for
these components. Finally, the applications take $3 x s$ and take $3 y s$ yield the prefixes of length three of the lists $x s$ and $y s$, respectively.

The evaluation of this let expression does not terminate if we use the original implementation of partition. At that time the implementation of partition inspected its whole argument, which is the infinite list [1..] in this example. In contrast, if we define partition by means of filter the evaluation of the expression above yields the pair $([1,3,5],[2,4,6])$. Here, the square brackets denote a list and each pair of consecutive elements of the list is separated by a comma. That is, a definition of partition by means of filter yields a list with the first three odd numbers and a list with the first three even numbers as this definition inspects only a small part of the infinite list [1..]. More precisely, the application take $3 x$ as well as take $3 y s$ only inspect the first three elements of $x s$ and $y s$, respectively. The application filter odd [1..] inspects the first five elements of [1..] to yield the list $[1,3,5]$. Moreover, the application filter (not $\circ$ odd ) [1..] inspects the first six elements of [1..] to yield the list $[2,4,6]$. Thus, both applications of filter inspect only a finite part of the infinite list.

The implementation of partition has been improved by introducing a so-called lazy pattern, which delays a check whether a value has a certain structure. Essentially, the definition of partition has to check whether the result of its recursive application is a pair to access the components of the pair. The original definition of partition accessed the components of the recursive application before yielding the first part of its result. To improve the behavior of partition with respect to strictness, the check for the pair structure of the recursive application has to be delayed.

Waldmann (2000) has already proposed to develop means to formally reason if a function can be improved by adding such a lazy pattern as well as tools that support the programmer in checking whether a function can be improved this way. One contribution of this thesis is a part-solution to his proposals. First, we present a criterion to check if a function is overly strict. This criterion is an answer to a more general question than the question whether a function can be improved by a lazy pattern. Furthermore, we present a tool called Sloth, which assists the programmer in checking whether a function is unnecessarily strict. In particular, the tool states that a function is overly strict in the case that it can be improved by introducing a lazy pattern. For example, in Section 6.6, using Sloth, we improve an implementation of the enumeration of elements of a tree in breadth-first order by introducing a lazy pattern. However, the tool only states whether a function is too strict and is not able to discover how a function can be improved. The tool, in particular, does not state whether a function can be improved by introducing a lazy pattern.

As another example for the interest in less strict implementations, let us consider the standard function permutations, which takes a list and yields a list of all permutations of the argument. The current definition of permutations emerged from a discussion initiated by van Laarhoven (2007).

Via a thorough refinement process the contributors of the discussion have defined a quite efficient implementation of permutations that, moreover, is less strict than the naive implementation they first came up with. More precisely, they have specified how little strict they expected an implementation of permutations to be. Intuitively, they stated that permutations is supposed to enumerate $n$ ! permutations, each of length $n$, by inspecting at most the first $n$ elements of its argument list.

This property can be considered as a kind of generalization of the concept of online behavior by Pippenger (1997). A function that takes a list and yields a list provides on-line behavior if the function inspects at most the first $n$ elements of its argument to yield $n$ elements of the result. The property of permutations, stated above, in fact, generalizes this idea to a function that yields a list of lists as result and to the special case of a function that enumerates permutations.

As a side note, functions with on-line behavior play an important role with respect to the expressiveness of pure functional programs with respect to complexity. Pippenger (1997) has shown that there are certain functions with on-line behavior that have a linear complexity in a strict functional language with destructive updates - he considers LISP - while in a strict language without destructive updates we always get an additional logarithmic factor. Bird et al. (1997) have provided an implementation of the same function in a lazy functional language that has linear complexity, too. That is, a lazy purely functional programming language is in a certain sense more expressive than a strict purely functional programming language. In this context the notion pure refers to the absence of destructive updates. Bird et al. (1997) make use of a less strict matrix transposition in order to get the desired on-line behavior in the lazy language. We consider this function in more detail in Section 8.2 and observe that they use a more liberal notion of less-strict than it is used in this thesis.

The notion of boundedness in the context of pretty printers (Wadler 2003) can also be considered as a kind of special case of on-line behavior. In the area of functional programming a pretty printer is a program or a library that takes a document that is represented by a tree structure and converts it into a string. The output has to reflect the structure that is represented by the tree structure of the document. Additionally, a pretty printer takes a width that restricts the number of characters in the string before each line break. A pretty printer is supposed to be optimal in the sense that it uses the smallest number of lines with respect to the given width, that is, line breaks are only added if necessary. Furthermore, a pretty printer is called bounded if it uses a look-ahead of at most $w$ characters to pretty print a document with width $w$.

That is, boundedness aims for an implementation of pretty printers that inspects only a small part of its argument to yield a certain part of its result in the same way as the property of permutations and the more general term of on-line behavior. In other words, all these concepts aim for implementations whose strictness does not exceed a certain degree. In contrast, in this thesis we aim for implementations that are as little strict as possible. A function that is as little strict as possible is called minimally strict. For example, we can observe that permutations is actually minimally strict, at least for lists up to six element, by using the tool that is presented in this thesiss. For most functions we can only make a statement whether the function is minimally strict for a restricted number of arguments where we use a notion of size for the restriction. Sloth is only able to verify that permutations is as little strict as possible for lists up to six elements as the result of permutations has factorial many elements in the number of elements in the argument list. Thus, checking the result of each application is quite expensive simply because of the number of elements.

Finally, there are two case studies that use a heap profiling tool for a lazy functional programming language to improve the memory usage of a program. In both
cases, the memory usage of one of the considered functions is improved by using a less strict implementation. First, Runciman and Wakeling (1993a) have applied their profiling tool to check the memory usage of a compiler for the programming language Lazy ML, a lazy dialect of the strict functional programming language ML. In this case study they present an example of a function whose memory usage can be improved by using a less strict definition. However, they do not provide enough background information to relate this example to the results presented in this thesis. Furthermore, Runciman and Wakeling (1993b) have used the profiling tool to investigate a program that takes a propositional formula and yields a clausal normal form of this formula. We investigate this example in detail in Section 8.2 and observe that their notion of less-strict goes beyond the notion of less-strict considered in this thesis.

The related works considered so far make use of concepts that are similar to the idea of minimally strict functions. Besides these works the most important related work is a draft paper by Chitil (2006) ${ }^{1}$. He has presented the idea to automatically check whether a function is as little strict as possible. Furthermore, he has presented the idea of how to generally observe that a function is unnecessarily strict. The tool presented in this thesis basically still applies the same idea. Chitil (2006) has even presented a tool called StrictCheck that checks whether a function is as little strict as possible. However, StrictCheck proposes functions that are only implementable using parallel features like the ambiguous choice operator by McCarthy (1961). We can model this operator in Haskell using non-pure features, which functional programmers want to avoid most of the time. Ambiguous choice and non-pure features break referential transparency in the sense that we can no longer replace equals by equals.

Finally, Coutts et al. (2007), as a side-product, have developed a tool that is related to the concept of minimally strict functions and Sloth. They have developed a library for stream processing and implemented standard list functions on basis of this library. Similar to the partition example they have used specifications from the language report to check whether their implementations agree with the standard definitions of these functions. To compare the functions they have implemented a tool for checking whether two functions agree for all possible inputs up to a specific size similar to SmallCheck by Runciman et al. (2008). However, in contrast to SmallCheck they also consider that an argument might contain an error or a non-terminating expression. This approach circumvents the problem of parallel functions as two functions are compared with each other. Thus, if the function that acts as specification is not parallel, the tool will not propose a parallel implementation. Nevertheless, in most cases we do not have a specification that can be used to check a function and, furthermore, we still do not know whether the function can be improved with respect to strictness as the specification might be unnecessarily strict as well. For example, in Section 3.2 we observe that the implementation of the function intersperse as it is given in the language report is unnecessarily strict.

The rest of this thesis is organized as follows.

[^0]- Chapter 2 introduces the main features of the lazy, purely functional programming language Haskell (Peyton Jones 2003). Furthermore, it provides a denotational semantics for a simple non-strict, first-order, functional programming language without polymorphism. This language is used as basis for a mathematical model of minimal strictness in Chapter 4.
- Chapter 3 illustrates the concept of non-strict evaluation and presents a practical example that backs the interest in less strict functions. More precisely, it presents an application whose memory usage can be improved by a factor of 20,000 by using a less strict implementation of a standard Haskell function, namely above-mentioned intersperse.
- Chapter 4 first revisits the approach by Chitil (2006). Furthermore, it presents a criterion to check whether a function is minimally strict and rule out parallel implementations by employing the notion of sequentiality. This chapter as well as the underlying denotational semantics only consider monomorphic functions. Finally, we prove that the presented criterion is necessary and sufficient for the existence of a less strict, sequential function.
- Chapter 5 presents the basics of a light-weight tool, that is, a Haskell library, called Sloth, that is based on the mathematical model of Chapter 4. In particular, the chapter illustrates certain approximations that are applied to gain a light-weight tool. Parts of the results of this chapter have already been presented elsewhere (Christiansen 2011).
- Chapter 6 generalizes the concept of minimally strict functions to polymorphic functions. We show how we can check whether a polymorphic function is minimally strict by checking a specific monomorphic instance of the polymorphic function. Furthermore, we prove that we can reduce the number of test cases of a polymorphic function significantly by employing this approach. The results of this chapter have already been presented (Christiansen and Seidel 2011).
- Chapter 7 presents some case studies of checking whether functions are minimally strict. More precisely, we consider Peano arithmetic, arithmetic for a representation of binary numbers by Braßel et al. (2008), and a Haskell library called split (Yorgey 2011). Besides, we also present an approach to derive less strict functions in certain cases and exemplarily discuss the connection between less strictness and space usage by means of the function reverse that reverses the order of the elements of a list.

Finally, we want to make some organizational remarks. All proofs that are omitted in Chapter 4 and Chapter 6 are presented in Appendix A and Appendix B, respectively. For all run-time and memory usage measurements we have used an Apple MacBook Pro 2.3 GHz with an Intel Core i5 and 4GB RAM. When we benchmark Haskell programs, we use the Glasgow Haskell Compiler (GHC) version 7.0.3 in single core mode. At last, we want to make a remark about mathematical rigorousness. The mathematical model in Chapter 4 is based on the denotational semantics presented in Section 2.2. In contrast, Chapter 6 as well as Chapter 7 use a more
loose approach and do not strictly separate syntax and semantics. This approach is quite common in the context of functional programming as the semantics of a functional language is very closely related to the syntax of the language. We only have to take extra care to consider the cases that an expression yields an error or does not terminate. Chapter 6 gives more details about this approach to reasoning.

1 Introduction

## 2 Preliminaries

This section presents some preliminaries for the remainder of this thesis. The first part of this section presents the main features of the functional programming language Haskell (Peyton Jones 2003). The reader who is familiar with Haskell may safely skip Section 2.1. In Section 7.3 and Section 8.2.2 we also give an outlook to applications of the presented approach to the functional logic programming language Curry (Hanus 2006), but we do not introduce the features of this language till we need them.
Section 2.2 presents a simplified functional programming language and provides a standard denotational model for this language. This denotational model is used in Chapter 4 as the basis for a mathematical model of minimally strict functions. The reader who is familiar with denotational semantics of non-strict functional languages may only inspect the syntax of the considered language presented in Figure 2.2.1 and note that we do not use arbitrary complete partial orders as domains but bounded complete, algebraic cpos. Definitions of these concepts are found in Section 2.2.

### 2.1 Introduction to Haskell

While in an imperative programming language the model of programming is based on destructively modifying the contents of memory cells, in a functional programming language the model of programming is based on the definition and application of functions. In other words, while in an imperative programming language a variable is a name for a location in the memory, in a functional programming language a variable is a shortname for a more complex expression. That is, while we can change the value that is assigned to a variable in an imperative language we cannot change it in a functional language. In this section we introduce the main features of the functional programming language Haskell. We start by showing how data is represented in Haskell, more precisely, we introduce the concept of algebraic data types.

### 2.1.1 Algebraic Data Types

A data type definition is introduced by the keyword data followed by the name of the algebraic data type, which has to start with a capital letter. One of the simplest algebraic data types, but a quite essential one is the type of Boolean values Bool.

$$
\text { data Bool }=\text { False } \mid \text { True }
$$

In the case of the data type of Boolean values the type is called Bool. The name is followed by an equality sign and an enumeration of the constructors of the type,

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separated by vertical bars. In the case of Bool the constructors are called False and True. The name of a constructor has to start with a capital letter as well. A constructor is used to construct a value of the corresponding type. That is, the term False as well as the term True construct a corresponding value of type Bool.

While the constructors of Bool are both constants we can define constructors that take additional arguments. For example, the following data type Partial defines a constant None and an unary constructor Some, which takes an integer as argument.

$$
\text { data } \text { Partial }=\text { None } \mid \text { Some Int }
$$

Besides algebraic data types, Haskell provides primitive data types, for example, Int for integers, Integer for arbitrary precision integers, Float and Double for floating point numbers, and Char for characters. To construct a value of type Partial we can either use None or a term like Some 42, for example. That is, we apply a constructor to an argument by juxtaposition of the constructor name and the argument.

A data type like Partial can be used to represent the result of a function if the function is only defined for some inputs. For example, consider a function that takes some kind of mapping from integers to integers and looks up a key in this mapping. If the mapping does not contain a binding for a key, the result of the function is None. If the mapping contains a binding, the result is Some $i$ where $i$ is the integer that is associated with the provided key. Obviously, we potentially also need a data type like Partial where the Some constructor contains a Boolean value, for example, instead of an integer. As this data type would be very similar to Partial, Haskell provides a mechanism to abstract over the concrete data type whose values may be used for arguments of the Some constructor.

Instead of the data type Partial for integers and another data type for Boolean values, Haskell provides the following data type, called Maybe.

$$
\text { data Maybe } \alpha=\text { Nothing } \mid \text { Just } \alpha
$$

The $\alpha$ in this equation is called type variable and denotes that the type Maybe takes a type as argument. Types that take a type as argument are called type constructors. Instead of the type Partial we can as well use the type Maybe Int, that is, we apply the type constructor Maybe to the type Int. This way we instantiate the type variable $\alpha$ with the type Int and get a type Maybe Int, whose values are Just applied to some integer and Nothing. For example, the term Just 42 is a value of type Maybe Int. In contrast, the term Just False constructs a value of type Maybe Bool.

Besides the simple data type definitions, considered so far, we can define more complex data types by employing recursion. The following algebraic data type defines a type of ordered collections of values of equal type, called lists.

$$
\text { data List } \alpha=\text { Nil } \mid \text { Cons } \alpha(\text { List } \alpha)
$$

For an arbitrary type $\tau$ we can construct an empty list of type List $\tau$ by using the term Nil. In the following we use the names $\alpha, \beta, \gamma$ for type variables and $\tau, \tau_{1}$, $\tau_{2}, \tau_{3}$ as meta-variables for concrete types. We can construct a non-empty list of type List $\tau$ by using the constructor Cons that takes a value of type $\tau$ and a list with
elements of type $\tau$ as second argument. For example, the term Cons 1 Nil constructs a list of integers, that is, a list of type List Int, that contains one element, namely 1. Furthermore, the term Cons 1 (Cons 2 Nil) constructs a list with the elements 1 and 2. Note that a list like Cons 1 (Cons False Nil) is not a value of any list type. The list Cons False Nil is an element of type List Bool. Therefore, Cons applied to this value may only take a value of type Bool as first argument as the constructor Cons states that its first argument has the same type as the elements of the list of its second argument.

As lists are a very commonly used data type in Haskell, the actual syntax of the list data type is prettier than the one above. The following definition is not valid Haskell syntax, but it illustrates the syntax of lists in Haskell. The constructors of this data type are [] and ":", where the latter one is used infix.

$$
\operatorname{data}[\alpha]=[] \mid \alpha:[\alpha]
$$

We can construct the empty list by [] and a list of type [Int] with the element 1 by 1 : []. Furthermore, a list with 1 and 2 is constructed by $1: 2:$ []. Note that the infix list constructor ":" is right associative, that is, the term $1: 2:[]$ stands for $1:(2:[])$. There is also a pretty syntax for the construction of lists. For example, the term [False, True] denotes the list False: True: [] and the term [1,2,3,4] denotes the list $1: 2: 3: 4:[]$.

Type constructors are not restricted to a single argument. The following algebraic data type defines a type of pairs by using a two-ary type constructor. Note that the name of the type may agree with the name of a constructor.

$$
\text { data Pair } \alpha \beta=\text { Pair } \alpha \beta
$$

The constructor Pair takes two arguments. The first argument is a value of the type that is passed as first argument to the type constructor Pair. The second argument of the constructor Pair is a value of the type that is passed as second argument to the type constructor Pair. For example, the term Pair False 42 constructs a pair of type Pair Bool Int.

Haskell uses a pretty syntax for pairs, too. In contrast to the list data type, pairs are not constructed by an infix constructor but by a mixfix constructor. Again, the following definition is not valid Haskell syntax but illustrates the syntax of pairs in Haskell.

$$
\text { data }(\alpha, \beta)=(\alpha, \beta)
$$

The pair that we have considered above is constructed by (False, 42) and it is an element of the type (Bool, Int).

### 2.1.2 Functions

As mentioned before, functions are quite essential in a functional programming language. In Haskell, functions are defined by means of pattern matching, where a pattern is a kind of inverse of a constructor. While we can use a constructor to assemble data, we can use a pattern to disassemble it again. As an example, we consider the following function, which disassembles a value of type Maybe Int.

```
fromJust (Just \(x)=x\)
```

A function definition starts with the name of the function, in this case fromJust. The first character of a function name has to be lower case to distinguish function names from constructor and type names. The name of a function is followed by a pattern. In the case of fromJust the pattern matches if the argument is a value constructed by using the constructor Just. All variables used in a pattern are bound to the corresponding value in the argument, the function is applied to. Note that we can distinguish variables from constructors because variable names start with a lower case letter. For example, if we apply fromJust to Just 42, then the variable $x$ is bound to 42 and the result of this application is 42 as well.

The application of a function to an argument, like the application of a constructor to an argument, is denoted by juxtaposition of function name and argument. That is, fromJust (Just 42) denotes the application of fromJust to the argument Just 42. If we apply fromJust to Nothing, we get a run-time error as fromJust does not provide a rule for this case.

To define a function that matches more than one constructor, we can use multiple rules, written below each other. For example, the following function defines the Boolean negation by using two rules, one for the case that the argument is False and one for the case that the argument is True.

$$
\begin{aligned}
\text { not False } & =\text { True } \\
\text { not True } & =\text { False }
\end{aligned}
$$

The rules of a function are processed from top to bottom. Furthermore, variables are also valid patterns. For example, we can as well define the Boolean negation as follows.

$$
\begin{aligned}
\text { not False } & =\text { True } \\
\text { not } b & =\text { False }
\end{aligned}
$$

Here the first rule matches if the argument is False and the second rule matches in all other cases, that is, if the argument is True. If a variable that is introduced by a pattern is not used on the right-hand side of the equality sign, as it is the case for the second rule of not, we can use an underscore instead of the variable in the pattern as well.

A function can also take multiple arguments where the arguments are separated by whitespaces. The following definition of andL defines the Boolean conjunction in Haskell.

```
andL False \(=\) False
andL True \(b=b\)
```

If the first argument of andL is False, the result of the function is False as well, independent of the second argument. If the first argument is True, then andL yields its second argument.

By means of this simple example we can already observe that there are often several ways to define the same basic behavior. For example, we can define the following implementation of the Boolean conjunction, called and $R$, which performs pattern matching on its second argument.

$$
\begin{aligned}
\text { and } R_{-} \text {False } & =\text { False } \\
\text { and } b \text { True } & =b
\end{aligned}
$$

Besides, there is one implementation of the Boolean conjunction, called and, that explicitly lists all cases as follows.

$$
\begin{aligned}
\text { and False False } & =\text { False } \\
\text { and False True } & =\text { False } \\
\text { and True False } & =\text { False } \\
\text { and True True } & =\text { True }
\end{aligned}
$$

Later we will observe that all these ways to define a Boolean conjunction lead to different behavior with respect to strictness.

The predefined Boolean conjunction in Haskell is an infix operator called " \&\&". The definition of an infix operator does not follow the general rule of a function definition. In contrast to a function definition, the name of an infix operator is written between its two arguments.

$$
\begin{aligned}
& \text { False } \& \&-=\text { False } \\
& \text { True } \& \& b=b
\end{aligned}
$$

If we want to use an infix operator in the way as we can use a two-ary function, we enclose its name in parentheses. For example, this way we can define the infix operator ( \&\&) by first stating its name.
(\&\&) False $=$ False
(\&\&) True $b=b$
In the same way as we can transform an infix operator into a normal, two-ary function by enclosing it in parentheses we can as well use a two-ary function as infix operator by enclosing it in quotes. For example, the term False 'and' True applies the two-ary function and to the arguments False and True.

Besides pattern matching in function rules, we can use pattern matching in socalled case expressions. By using a case expression a function can behave differently depending on the value of an expression in the same way as a function rule specifies the behavior of a function depending on the argument of the function. The following definition is an alternative to the definition of the negation function by means of several rules.

```
not b=
    case b of
        False }->\mathrm{ True
        True }->\mathrm{ False
```


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The keyword case is followed by the expression over which we branch, in this case the variable $b$. This expression is also called the scrutinee of the case expression and it is itself followed by the keyword of. The keyword of is followed by pairs of pattern and expression, also called branches, where each pattern is separated by $\rightarrow$ from the corresponding expression ${ }^{1}$. Like function rules, the patterns of a case expression determine the right-hand side that is used, depending on the value of the scrutinee.

The branches of a case expression can be separated by semicolons and enclosed in braces to allow unambiguous definitions of cascaded case expressions. Alternatively, as used in the example above, we can use a so-called offside rule (Landin 1966). That is, the leftmost character of the first pattern introduces a column that is used to identify the next pattern. All tokens that begin in this column are considered as patterns and start a new branch. All tokens that begin to the right of this column are still assigned to the previous rule while all tokens that begin to the left of this column close the definition of the branches of the case expression. This way we can leave out the braces and semicolons in the definition of a case expression.

Haskell also provides an if-then-else expression that can be considered as syntactic sugar for a case expression that branches over a Boolean value. More precisely, the following semantic equivalence holds where we use braces and semicolons to explicitly separate branches instead of using the offside rule.

$$
\text { if } b \text { then } e_{1} \text { else } e_{2} \equiv \text { case } b \text { of }\left\{\text { True } \rightarrow e_{1} ; \text { False } \rightarrow e_{2}\right\}
$$

Note that here and in the following we always use the symbol $\equiv$ for semantic equivalence. In Section 2.2 we provide a formal definition of semantic equivalence for a simple functional language, which can be considered as a subset of Haskell.

Before we go on, we introduce two other syntactical constructs that are often used to define functions in Haskell. These constructs are called guards and where clause. Instead of using an equality sign in the definition of a function we can use a vertical bar to introduce a definition by means of guards. Guards are Boolean expressions that, similar to a case expression, are used to branch over the value of an expression. The vertical bar is followed by a Boolean expression and the expression is followed by an equality sign and the corresponding right-hand side. For example, the following definition implements the signum function that takes an arbitrary integer and yields -1 if the integer is negative, 0 if the integer is 0 , and 1 if the integer is positive.

```
signum \(i\)
    \(\mid i<0=-1\)
    \(i==0 \quad=0\)
    otherwise \(=1\)
```

Guards are processed from top to bottom. The operator $(<)$ is a two-ary Boolean predicate that checks whether an integer is strictly smaller than another integer and the operator $(==)$ checks whether two integers are equal. In Section 2.1.6 we will

[^1]observe that the operator $(==)$ actually does not only work for integers but for a much larger class of values. The constant otherwise is synonym for the Boolean constant True. Therefore, signum yields -1 if $i$ is smaller than zero, 0 if $i$ is equal to zero and 1 otherwise.

Finally, we want to present so-called where clauses. A where clause introduces local function and variable definitions. We can add a where clause to each rule of a function. The variables introduced by the patterns of the rule are visible in the righthand sides of the definitions in a where clause. Functions and variables defined in a where clause are only visible in the right-hand side of the corresponding rule. For example, consider the following implementation of the reverse function that uses an accumulating parameter to achieve a linear complexity.

```
reverse \(=\) rev []
    where
        revexs [] \(=x s\)
        rev \(x s(y: y s)=\operatorname{rev}(y: x s) y s\)
```

We define a local function called rev in the where clause. The function rev is neither visible in additional rules of reverse nor in other top-level definitions. By top-level definitions we denote function and variable definitions that are not defined within a where clause or a let expression.

Note that we call it a where clause and not an expression as it does not play the same role as an expression. For example, a where clause cannot be the argument of an application while an expression can. However, there is also a counterpart of the where clause in the family of expressions, called let expression. A let expression consists of two keywords let and in. Between these keywords we can define functions and variables in the same way as we would define them using a where clause. These functions and variables are only visible in the expression that follows the keyword in. We can define the reverse function by employing a let expression as follows.

```
reverse =
    let rev xs [] = xs
        rev xs (y:ys)=rev (y:xs) ys
    in
    rev []
```

Finally, note that the where clause as well as the let expression make use of offside rules similar to the offside rule of the case expression. We abstain from a more thorough explanation as we do not need a more precise understanding of offside rules in the following and refer the interested reader to the Haskell language report (Peyton Jones 2003).

### 2.1.3 Types

So far we have stated that each constructor belongs to a certain type. That is, we can assign a type to every constructor. However, we can even assign a type to every function and, moreover, to every expression in a Haskell program.

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Haskell has a strong, static type system with type inference. The type system is said to be strong because there will be no type errors at run-time if we run a welltyped program. Sometimes people also refer to a language as strongly typed if there are no implicit coercions. For example, the expression False +2 is not well-typed in Haskell because $(+)$ expects two values of type Int, and False is of type Bool. Note that there are typed languages that use an implicit coercion to apply $(+)$ to False and 2 , for example, by considering the Boolean value False as the integer 0 . The type system of Haskell is said to be static because the compiler determines at compile time whether a program is well-typed and rejects the program if it is not. In contrast, in a dynamically typed language like python we might get a type error when we execute the program. Finally, type inference means that the compiler is able to determine the type of an arbitrary expression (Damas and Milner 1982). Therefore, we do not have to annotate any types at all and the compiler is still able to check whether the program is well-typed. Nevertheless, for documentation purposes, we annotate a type to every top-level Haskell definition in the following.

We annotate a type to an expression by separating the expression and the type by two colons. For example, the term False :: Bool annotates the type Bool to the constructor False. If we annotate a wrong type, for example, False :: Int, the compiler reports a type error.

We annotate a type to a function by preceding the function definition with the function name, two colons, and the type of the function. We annotate the type Bool $\rightarrow$ Bool to the function not as follows.

$$
\begin{aligned}
& \text { not }:: \text { Bool } \rightarrow \text { Bool } \\
& \text { not False }=\text { True } \\
& \text { not True }=\text { False }
\end{aligned}
$$

Here the right-arrow $\rightarrow$ is a two-ary type constructor that constructs a function type. That is, not is a function that takes a Boolean value as argument and yields a Boolean value as result. Consider the two-ary Boolean conjunction (\&\&) as another example.

$$
\begin{aligned}
& (\& \&):: \text { Bool } \rightarrow \text { Bool } \rightarrow \text { Bool } \\
& \text { False \&\& }=\text { False } \\
& \text { True } \& \& b=b
\end{aligned}
$$

The type of this function is Bool $\rightarrow$ Bool $\rightarrow$ Bool. More precisely, if a function takes more than one argument the additional arguments are separated by $\rightarrow$ from each other. For example, a function that takes three arguments of types $\tau_{1}, \tau_{2}$, and $\tau_{3}$ has the type $\tau_{1} \rightarrow \tau_{2} \rightarrow \tau_{3} \rightarrow \tau_{4}$ where $\tau_{4}$ is the result type of the function. Note that the function arrow is right-associative, that is, $\tau_{1} \rightarrow \tau_{2} \rightarrow \tau_{3}$ stands for $\tau_{1} \rightarrow$ $\left(\tau_{2} \rightarrow \tau_{3}\right)$. Because the function arrow is right-associative, function application is left-associative, thus, $f e_{1} e_{2} e_{3}$ stands for $\left(\left(f e_{1}\right) e_{2}\right) e_{3}$.

### 2.1.4 Parametric Polymorphism

In fact, we have already introduced parametric polymorphism when we have talked about algebraic data types. Parametric polymorphism means that a data type or a
function is parametric over a type. For example, the list data type is parametric over the type of the elements of the list. However, parametric polymorphism is not restricted to data types but can also be used for function definitions. In the following we present several examples of polymorphic functions.

Many functions, especially functions over collection types like lists or trees, behave equally for different element types. For example, counting the number of elements in a list does not depend on the type of the elements in the list. Consider the following definition of the standard Haskell function length that takes a list and yields its length. Here and in the following, to avoid parentheses, we employ the convention that a function, written prefix, has a higher precedence than an infix operator. Hence, the expression $1+$ length $x s$ is short for $1+($ length $x s)$.

$$
\begin{aligned}
\text { length }[] & =0 \\
\text { length }(-: x s) & =1+\text { length } x s
\end{aligned}
$$

Obviously, the function length has the type [Bool] $\rightarrow$ Int as it can calculate the length of a list of Boolean values. However, it seems equally valid to assign the type $[$ Int $] \rightarrow$ Int to length as it can also calculate the length of a list of integers. And, indeed, both types are valid types for length. The compiler does not report a type error if we annotate one of these types to the function length. Nonetheless, there is a more general type in the sense that we can apply the length function to more arguments without risking the type checker to complain. In the same way as we have used type variables to express that a data type is parametric, we use type variables to express that length calculates the length of a list of an arbitrary element type. More precisely, we can annotate the type $[\alpha] \rightarrow$ Int to length.

In Haskell the type of a function is always implicitly quantified over all type variables that occur in it. Furthermore, the variables are quantified at the beginning of the type. For example, the type $[\alpha] \rightarrow$ Int stands for the type $\forall \alpha .[\alpha] \rightarrow$ Int. It is just a question of convenience to leave out the quantifier. Though, note that it is quite important that all variables are quantified at the beginning of a type. For example, the type $\forall \alpha .[\alpha] \rightarrow$ Int and the type $[\forall \alpha . \alpha] \rightarrow$ Int mean totally different things. The latter is also referred to as a higher-rank type (Odersky and Läufer 1996). The former type denotes a function that takes a list of an arbitrary type $\tau$ and yields an integer or, more precisely, we can choose some type $\tau$ and get a function of type $[\tau] \rightarrow$ Int. In contrast, the latter type denotes a function that takes a list of polymorphic values as argument and yields an integer. That is, if we assign the type $[\forall \alpha . \alpha] \rightarrow$ Int to length we can only apply it to lists where every element has the type $\forall \alpha . \alpha$. For example, this is not the case for a list of Boolean values. Although Haskell provides extensions to higher-rank types (Peyton Jones et al. 2007), in the following we only consider rank-1 polymorphism. In other words, all type variables of function types are implicitly quantified at the beginning of the type definition.

A more sophisticated example of a polymorphic function is the list concatenation $(+)$. In this case we employ a type variable to state that the element type of the result list is the same as the element type of the argument list. Furthermore, as both arguments are of type [ $\alpha$ ], we know that both argument lists have to have the same element type, so we cannot concatenate a list of Boolean values with a list of integers.

$$
\begin{aligned}
& (+)::[\alpha] \rightarrow[\alpha] \rightarrow[\alpha] \\
& {[] \quad+y s=y s} \\
& (x: x s)+y s=x:(x s+y s)
\end{aligned}
$$

Note that we cannot define a concatenation function that takes one list of type $[\alpha]$ and one list of type $[\beta]$ because of the definition of the list data type. More precisely, the application of the list constructor (:) to $x$ and the result of $x s+y s$ in the second rule of $(+)$ implies that $x$ has the same type as the elements of the result of $(+)$. Furthermore, the first rule of $(+)$ implies that the list $y s$ has the same type as the result of (\#).

We can as well use multiple type variables in a function type to specify more complex relationships. For example, the following function takes two lists of potentially different element types and combines these lists point-wise to a list of pairs.

```
zip :: [\alpha]->[\beta]->[(\alpha,\beta)]
zip}(x:xs)(y:ys)=(x,y):zipxs y
zip_ _ = []
```

By employing type variables we express that the elements of the first list have the same type as the values of the first component of the pair and the elements of the second list have the same type as the values of the second component of the pair. Note that we could as well annotate the type $[\alpha] \rightarrow[\alpha] \rightarrow[(\alpha, \alpha)]$ to zip, but in this case the function is less general. For example, while we can apply zip to the lists [False] and [42] this application would not be well-typed if we annotate the less general type because the less general type only allows argument lists of equal type.

Finally, we want to introduce a function that, in particular, is intensively used in Chapter 6. The list indexing function (!!) :: $[\alpha] \rightarrow$ Int $\rightarrow \alpha$ takes a list and an index in the list (starting with zero) and yields the element at the corresponding position.

$$
\begin{aligned}
& (!!)::[\alpha] \rightarrow \text { Int } \rightarrow \alpha \\
& {[] \quad!!\quad=\text { indexError }} \\
& (x: x s)!!i= \\
& \mid i<0 \quad=\text { indexError } \\
& \mid i==0=x \\
& \mid \text { otherwise }=x s!!(i-1)
\end{aligned}
$$

If the list is empty or the index is smaller than zero, the function yields a run-time error denoted by indexError. Using a guard we distinguish the cases that the index $i$ is smaller zero, equal to zero, or greater than zero. In the second case the first element of the given list is the element we are looking for and in the third case we have to project to position $i-1$ of the list without the first element, denoted by $x s$.

### 2.1.5 Higher-Order

In Haskell, functions are first class citizens. That means a function can be used, like any other data type, as argument or result of a function or even as argument of a constructor.

As we have observed before, the function arrow $\rightarrow$ is right associative. Therefore, the type of a two-ary function $\tau_{1} \rightarrow \tau_{2} \rightarrow \tau_{3}$, in fact, stands for the type $\tau_{1} \rightarrow$ $\left(\tau_{2} \rightarrow \tau_{3}\right)$. That is, if we apply the function to a value of type $\tau_{1}$ we get a function of type $\tau_{2} \rightarrow \tau_{3}$. We can even partially apply any function that takes more than one argument. If we apply a function that takes $n$ arguments to $m$ arguments, where $m$ is smaller than $n$, we get a function that takes another $n-m$ arguments. For example, the term and True denotes a partial application of the function and to the argument True. More precisely, the term and True denotes a function of type Bool $\rightarrow$ Bool that is semantically equivalent to the identity function of type Bool $\rightarrow$ Bool. In the same way we can apply a function of type $\tau_{1} \rightarrow \tau_{2} \rightarrow \tau_{3} \rightarrow \tau_{4}$ to a value of type $\tau_{1}$ and get a function of type $\tau_{2} \rightarrow \tau_{3} \rightarrow \tau_{4}$. If we, again, apply the resulting function to a value of type $\tau_{2}$, we get a function of type $\tau_{3} \rightarrow \tau_{4}$.

To partially apply an infix operator we can use so-called left and right sections. We enclose an infix operator together with one argument in parentheses and indicate by the position of the argument to the left or to the right of the infix operator if we partially apply the operator to its first or its second argument. For example, the expressions (True \&\&) and (\&\& True) partially apply (\&\&) to True and the resulting function still awaits its second and first argument, respectively.

As mentioned before, functions cannot only yield functions as result but also take functions as argument. Consider the following function definition that takes a predicate and a list and removes all elements from the list that do not satisfy the predicate. Here, the variable pred stands for a function of type $\tau \rightarrow$ Bool, and, therefore, we can apply it to the list element $x$. When we apply pred to $x$, we get a Boolean value as result. Thus, we can use a guard to branch over the result of this application. Note that we may not leave out the parentheses in the type of filter this time as the function arrow is right-associative.

$$
\begin{aligned}
& \text { filter }::(\alpha \rightarrow \text { Bool }) \rightarrow[\alpha] \rightarrow[\alpha] \\
& \text { filter_ } \quad[]=[] \\
& \text { filter pred }(x: x \mathrm{~s}) \\
& \quad \mid \text { pred } x=x: \text { filter pred } x \mathrm{~s} \\
& \quad \mid \text { otherwise }=\text { filter pred } x \mathrm{~s}
\end{aligned}
$$

A mathematically natural higher-order function is the function composition (o). The function composition takes a function of type $\tau_{2} \rightarrow \tau_{3}$, a function of type $\tau_{1} \rightarrow \tau_{2}$ and a value of type $\tau_{1}$ and applies these functions one after the other to the value.

$$
\begin{aligned}
& (\circ)::(\beta \rightarrow \gamma) \rightarrow(\alpha \rightarrow \beta) \rightarrow \alpha \rightarrow \gamma \\
& (f \circ g) x=f(g x)
\end{aligned}
$$

Note that this definition shows how we can define an infix operator that takes more than two arguments by using parentheses. As an example, the term not o not is the identity function of type Bool $\rightarrow$ Bool as it takes a Boolean argument and applies not twice to this argument.

Higher-order functions are a means to define abstractions that can be employed in many contexts. The function map is one example for a frequently used abstraction

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of this kind. It takes a function as higher-order argument and applies this function point-wise to all elements of a list.

$$
\begin{aligned}
\operatorname{map}::(\alpha \rightarrow \beta) & \rightarrow[\alpha] \rightarrow[\beta] \\
\operatorname{map}-[] & =[] \\
\operatorname{map} f(x: x s) & =f x: \operatorname{map} f x s
\end{aligned}
$$

As an example, we can apply map to not and the list [False, True, True] and get the list [True, False, False] as result.

Higher-order functions can be used to generalize first-order functions. For example, by employing a higher-order function we can generalize the definition of zip, presented in the previous section. Instead of constructing pairs from the elements of the argument lists, we apply an arbitrary two-ary function to each pair of elements. This two-ary function is provided in the form of a higher-order argument.

$$
\begin{aligned}
& \text { zipWith :: }(\alpha \rightarrow \beta \rightarrow \gamma) \rightarrow[\alpha] \rightarrow[\beta] \rightarrow[\gamma] \\
& \text { zipWith } f(x: x s)(y: y s)=f \text { xy:zipWithf xs ys } \\
& \text { zipWith }_{--}-\quad=[]
\end{aligned}
$$

Note that we have zip $\equiv$ zipWith (, ) where (, ) denotes the two-ary function that takes two arguments and constructs a pair.

In the context of higher-order functions, it is often useful to define a function without having to introduce a name for the function. Therefore, Haskell provides the possibility to define anonymous functions. An anonymous function is introduced by the symbol $\lambda$ followed by patterns that are separated by whitespaces. ${ }^{2}$ The patterns are followed by a right arrow and the right-hand side of the anonymous function. For example, the expression $\lambda x x s \rightarrow x s+[x]$ defines an anonymous function that takes an element $x$ and a list $x s$ and appends the element at the end of the list. As an example of the application of an anonymous function, the application filter $(\lambda i \rightarrow(i>0) \& \&(i<10))$ defines a function on lists of integers that removes all elements from the list that are smaller than one or greater than nine. Note that filter takes over the elements of the list for which the predicate is satisfied.

By using higher-order functions we can even define control structures that usually are built-in in other languages. For example, the following higher-order function is a functional implementation of the while loop, which is known from imperative programming languages. The first argument of while is a function that takes the current "state" of the loop and yields a Boolean value that determines whether the loop is executed once more. The second argument of while is a state transformation that is applied in every execution of the body to get a new state from the old one.

```
while \(::(\alpha \rightarrow\) Bool \() \rightarrow(\alpha \rightarrow \alpha) \rightarrow \alpha \rightarrow \alpha\)
while pred trans \(x\)
    \(\mid\) pred \(x=\) while pred trans (trans \(x\) )
    \(\mid\) otherwise \(=x\)
```

As an example, we can use while $(\lambda(c,-) \rightarrow c>0)(\lambda(c, f) \rightarrow(c-1, c * f))(5,1)$ to calculate the factorial of 5 . More precisely, this application yields the pair $(0,120)$

[^2]where the first component represents a counter that is decreased in every execution of the loop body while the second component is the factorial of the initial counter, in our example 5 . The state transformation $(\lambda(c, f) \rightarrow(c-1, c * f))$ takes a counter and the current factorial, multiplies the factorial with the current counter, and decreases the counter.

A last comment about higher-order functions is in order to illustrate their power. If we consider a language that does only provide primitive recursive functions, we get an expressive power that is beyond primitive recursion if we add higher-order functions to the language (Hutton 1999). For example, we can even define the Ackermann function in a language with primitive recursion and higher-order functions.

### 2.1.6 Type Classes

Type classes (Wadler and Blott 1989) are an implementation of ad-hoc polymorphism, in the context of object-oriented programming languages also called overloading, in Haskell. Like a parametric polymorphic function can be applied to arguments of different types, an ad-hoc polymorphic function can be applied to arguments of different types as well. However, while a parametric polymorphic function behaves equally no matter which instance of the polymorphic function we consider, an ad-hoc polymorphic function, in most cases, behaves differently for different argument types. That is, we can provide several implementations of the same function that behave differently for different types. When we apply an overloaded function, the types of the arguments determine the concrete implementation that is used. ${ }^{3}$

Let us consider a function that tests two values for equality. Obviously, we cannot provide an equality check that works for arguments of different types. For example, we would like to implement an infix operator $\left(==_{\text {Bool }}\right)$ that is able to check two Boolean values for equality as follows.

$$
\begin{aligned}
& \left(==_{\text {Bool }}\right):: \text { Bool } \rightarrow \text { Bool } \rightarrow \text { Bool } \\
& \text { False }==_{\text {Bool }} \text { False }=\text { True } \\
& \text { True }==_{\text {Bool }} \text { True }=\text { True } \\
& \quad==_{\text {Bool }}-\quad=\text { False }
\end{aligned}
$$

In the same way as we have defined a function that checks Booleans for equality, we would like to define the following infix operator $\left(==_{[\text {Bool }]}\right)$ that checks two lists of Boolean values for equality.

$$
\begin{aligned}
&\left(=={ }_{[\text {Bool }]}\right)::[\text { Bool }] \rightarrow[\text { Bool }] \rightarrow \text { Bool } \\
&==_{[\text {Bool }]}[] \quad \\
& {[] \quad \text { True } } \\
&(b 1: b s 1)==_{[\text {Bool }]}(b 2: b s 2) \\
&=\left(b 1==_{\text {Bool }} \text { b2) \&\& (bs1 }==_{[\text {Bool }]} \text { bs } 2\right) \\
&==_{[\text {Bool }]}-\quad=\text { False }
\end{aligned}
$$

In Haskell a type class called $E q$ is used to implement equality checks like $\left(==_{\text {Bool }}\right)$ and $\left(==_{[\text {Bool }]}\right)$ and use the same name for both functions. The predefined type class $E q$ provides a function $(==)$ that implements an equality check. ${ }^{4}$

[^3]\[

$$
\begin{aligned}
& \text { class } E q \alpha \text { where } \\
& \quad(==):: \alpha \rightarrow \alpha \rightarrow \text { Bool }
\end{aligned}
$$
\]

The keyword class starts the definition of a type class, followed by the name of the class. Like constructors and types, the names of type classes have to start with a capital letter. The name of the class is followed by a type variable that denotes the type that will be an instance of this type class. After the keyword where we enumerate the functions that have to be implemented to make a type an instance of this type class. In the case of the type class $E q$, a type $\tau$ becomes an instance of the class if we provide a function ( $==$ ) of type $\tau \rightarrow \tau \rightarrow$ Bool. For example, the following code implements an instance of $E q$ for the type Bool.

> instance Eq Bool where
> False $==$ False $=$ True
> True $=$ True $=$ True
> _ $\quad==$ _ False

To make a type an instance of a type class we use the keyword instance. The name of the type class is followed by the specific type that becomes an instance of the type class, in this case Bool. The type is followed by the keyword where and by implementations of the functions of the type class, in this case $(==)$.

If we reconsider the equality check for lists of Booleans, we observe a more general structure. We define the equality check of lists by testing the elements of the list point-wise for equality. The equality check for lists of integers will look very similar to the equality check for lists of Booleans. We just have to replace all occurrences of $\left(==_{\text {Bool }}\right)$ by an equality check for integers. That is, instead of using a specific instance to check the equality of the elements (in the case of $\left(==_{[\text {Bool }]}\right)$ the specific instance is $\left.\left(==_{\text {Bool }}\right)\right)$ we would like to check the elements with the equality check of the element type of the list. We can use a type class constraint to define an instance like this. Using a type class constraint we make all types that satisfy a certain precondition an instance of the type class. The following instance makes all types of the form $[\tau]$ an instance of $E q$, if the type $\tau$ is an instance of $E q$.

$$
\begin{aligned}
& \text { instance } E q \alpha \Rightarrow E q[\alpha] \text { where } \\
& \begin{aligned}
& {[] \quad }==[] \quad \\
&=\text { True } \\
&(x: x s)==(y: y s)=(x==y) \& \&(x s==y s) \\
&-\quad==- \\
&-=\text { False }
\end{aligned}
\end{aligned}
$$

The applications of ( $==$ ) on the right-hand side are, in fact, applications of different functions. More precisely, while the left application $x==y$ invokes a function of type $\tau \rightarrow \tau \rightarrow$ Bool, the application $x s==y s$ invokes a function of the type $[\tau] \rightarrow[\tau] \rightarrow$ Bool. Note that, although we assume that these functions implement equality checks on the corresponding types, in fact, we do not know anything about these functions except for their types and their name.

Type class constraints are also used to restrict the arguments of certain functions. For example, consider that we want to define a function that takes a value and a list and checks whether the value is an element of the list. We obviously need an
equality test on the element type of the list to implement this function. That is, we want to define a function of type $\alpha \rightarrow[\alpha] \rightarrow$ Bool but we do not want to instantiate $\alpha$ with arbitrary types $\tau$ but only with types that are instances of $E q$. We can express this circumstance by adding the constraint $E q \alpha$ to the front of the type signature. The following definition of the membership test uses the Boolean disjunction implemented by the infix operator $(\|)$.

$$
\begin{aligned}
& \text { elem }:: \text { Eq } \alpha \Rightarrow \alpha \rightarrow[\alpha] \rightarrow \text { Bool } \\
& \text { elem }-[]=\text { False } \\
& \text { elem } x(y: y s)=(x==y) \| \text { elem } x \text { ys }
\end{aligned}
$$

If we omit the type class constraint in the type of elem, the compiler even reports a type error. By the use of $(==)$, the compiler knows that $x$ and $y$ must be values of a type that is an instance of $E q$.

We cannot only define type classes whose instances are types, but we can also define type classes whose instances are type constructors. This kind of classes is called type constructor classes (Jones 1993). For example, the following type class Functor is a quite prominent type constructor class.

$$
\begin{aligned}
& \text { class Functor } \varphi \text { where } \\
& \quad \text { fmap }::(\alpha \rightarrow \beta) \rightarrow \varphi \alpha \rightarrow \varphi \beta
\end{aligned}
$$

To make a type $\tau$ an instance of Functor, the type has to be a type constructor because the variable $\varphi$ is applied to a type, in this case type variables $\alpha$ and $\beta$. The type class Functor provides a function called fmap, which is a generalization of map. The function fmap takes a function from some type to another type and a value of some type constructor applied to the argument type of the function. Typically, an instance of Functor is some kind of collection that is parametric over the type of the elements, for example, a list, a tree, or some abstract data type like a set. In this case fmap can be considered as a function that applies its argument $f$ to every member of the collection.

In the case of the list type constructor the function fmap is equivalent to the function map, presented before.

```
instance Functor [] where
    fmap_[] = []
    fmap f (x:xs)=fx:fmap fxs
```

Note that the token [] in the first line of the definition denotes a type constructor that takes the element type of a list as argument and constructs the type of lists with this element type. In contrast, the occurrences of the token [] in the second line of the definition denote the pattern and the constructor for the empty list, respectively.

We can also regard a tuple as a collection with respect to one of its components. That is, for some type $\tau$ we can define an instance of Functor for the partially applied type constructor $((,) \tau)$ as follows. In this case we have to provide a function fmap of type $(\alpha \rightarrow \beta) \rightarrow(\tau, \alpha) \rightarrow(\tau, \beta)$. The following code implements an instance of Functor for $((,) \tau)$ where $\tau$ is an arbitrary type.

```
instance Functor \(((,) \alpha)\) where
    \(\operatorname{fmap} f(x, y)=(x, f y)\)
```

As mentioned before, instances of Functor are typically collections, but they don't have to be. For example, consider the two-ary type constructor $\rightarrow$ that constructs function types. If we partially apply $\rightarrow$ to some type $\tau$, denoted by $((\rightarrow) \tau)$, the resulting unary type constructor takes the result type of the function type as argument. To make this type constructor an instance of the type class Functor, we have to provide an implementation of fmap of type $(\alpha \rightarrow \beta) \rightarrow(\tau \rightarrow \alpha) \rightarrow(\tau \rightarrow \beta)$. This type is an instance of the type of the function composition (o). Therefore, we can define an instance of Functor for a partially applied function type as follows.

$$
\begin{aligned}
& \text { instance Functor }((\rightarrow) \alpha) \text { where } \\
& \text { frap } f g=f \circ g
\end{aligned}
$$

Nobody restrains us from defining "silly" instances of Functor. For example, the following implementation is also a valid implementation of fmap for the list type constructor. This instance ignores its arguments and yields the empty list.

```
instance Functor [] where
    fmap _ _ = []
```

The type class Functor is inspired by a concept with the same name in category theory. There, a functor is a homomorphism, that is, a structure preserving mapping, between two categories. More precisely, a functor is a mapping $F$ from a category $C$ to a category $D$ that assigns every object $X \in C$ an object $F(X) \in D$ and every morphism $f: X \rightarrow Y \in C$ a morphism $F(f): F(X) \rightarrow F(Y) \in D$. Furthermore, $F$ has to satisfy the following laws where $\circ$ denotes morphism composition.

$$
\begin{gathered}
F(i d)=i d \\
F(f \circ g)=F(f) \circ F(g)
\end{gathered}
$$

In contrast, in Haskell the functor $F$ is split into two parts, the type constructor that maps an object (in the case of Haskell a type) to an object and the function fmap that maps a morphism (in the case of Haskell a function) to a morphism. Moreover, for all instances of Functor, the function fmap is supposed to satisfy the following laws.

$$
\begin{gathered}
\text { fmap id } \equiv i d \\
\text { fmap }(f \circ g) \equiv \text { fmap } f \circ \text { fmap } g
\end{gathered}
$$

For example, our "silly" list instance breaks the first law. However, note that Haskell does not check whether an instance of Functor satisfies these laws.

### 2.2 Denotation of a Simple Functional Language

In this section we define a denotational semantics for a simple, non-strict, first-order, monomorphic functional language. This semantics is used in Chapter 4 to present a


Figure 2.2.1: Syntax of a Simple First Order Functional Language
mathematical model of minimally strict functions. Figure 2.2.1 presents the syntax of the considered language. For simplicity we only consider Booleans and lists as data types, but the results in Chapter 4 hold for arbitrary algebraic first-order data types. The language presented here is similar to languages considered elsewhere. For example, Plotkin (1977) presents a simple functional language called PCF (programming language for computable functions). In contrast to the language considered here PCF does not provide lists but provides integers instead, and PCF uses an explicit fixpoint combinator while we use a system of recursive function definitions.

Here and in the following, by $\tau$ we denote a type that is either Bool or a list type whose argument type is a type of the form $\tau$ again. By $\sigma$ we denote tuple types of arbitrary arity whose components are types of the form $\tau$. Tuple types only occur as argument types of functions and constructors.

A program $P$ is a sequence of declarations $D$ where $\epsilon$ represents the empty program. Furthermore, we use an uncurried notation for functions, that is, a function $f$ takes a tuple $\left\langle e_{1}, \ldots, e_{n}\right\rangle$ as argument. We do not use Haskell's curried notation as we do not consider higher-order functions and some of the following definitions are simplified by an uncurried notation. Function application is denoted by juxtaposition, for example, $f\left\langle e_{1}, \ldots, e_{n}\right\rangle$. We identify $f\rangle$ with $f$ and $f\langle x\rangle$ with $f x$. Instead of using semicolons to separate type annotation and function definition as well as braces and semicolons to separate case branches, for simplicity, we employ Haskelllike offside rules (Landin 1966). For convenience, we add an expression undefined $_{\tau}$ to the considered language that denotes a run-time error and is used to model partially defined case expressions.

## 2 Preliminaries

We assume that the reader is familiar with the basic concepts of partially ordered sets and complete partial orders. More precisely, we consider the concepts of chaincomplete partial orders and directed set complete partial orders and use these terms synonymously as these two concepts are equivalent. In the following we use the abbreviation cpo for complete partial order. Furthermore, by $\Pi$ and $\downarrow$ we denote the greatest lower bound (infimum) and the least upper bound (supremum) with respect to a corresponding ordering denoted by $\sqsubseteq$ and $\sqsupseteq$, respectively.

Instead of complete partial orders we interpret types by bounded complete, algebraic complete partial orders, also called Scott domains. Scott et al. (1989) show that flat cpos are bounded complete and algebraic, and that all standard domain constructions - which we employ here - preserve these properties.

A cpo is bounded complete if all subsets that have an upper bound also have a least upper bound. A bounded complete cpo is also a complete meet-semilattice, that is, the infima of all non-empty subsets exist. We can prove this statement by defining the infimum of a set as the supremum of all lower bounds of the set. This supremum exists because the cpo is bounded complete.

To establish algebraic cpos, we have to introduce the concept of finite (sometimes also called compact) elements. We prefer the term finite here because in the considered setting the informal notion of finiteness like in "finite list", for example, coincides with the following definition of finite elements.

Definition 2.2.1 (Finite Element): An element $x$ of a cpo $(D, \sqsubseteq)$ is called finite if for all chains $\left\langle x_{i}\right\rangle_{i \in I}$ in $D$ with $x \sqsubseteq \bigsqcup_{i \in I} x_{i}$ there exists $i \in I$ such that $x \sqsubseteq x_{i}$.

For any set $S$ the powerset $2^{S}$ together with the subset ordering $\subseteq$ is a cpo. The smallest element of a powerset cpo is the empty set, and the supremum of a chain of sets is given by the union of these sets. In the powerset cpo the finite elements are exactly the finite sets.

To prove the first direction, consider an infinite set $X$. Furthermore, consider a chain $\left\langle X_{i}\right\rangle_{i \in \mathbb{N}}$ of finite subsets of $X$ such that $\left|X_{i}\right|=i$. The supremum of this chain is the infinite set $X$, namely, $\bigcup_{i \in \mathbb{N}} X_{i}=X$. As there is no set $X_{i}$ such that $X \subseteq X_{i}$, the infinite set $X$ is not a finite element of $\left(2^{S}, \subseteq\right)$.

To prove the missing implication we proceed as follows. If $X$ is not finite, then there exists a chain $\left\langle X_{i}\right\rangle_{i \in I}$ such that $X \subseteq \bigcup_{i \in I} X_{i}$, and for all $i \in I$ we have $X \nsubseteq X_{i}$. If the index-set $I$ was finite, then the chain $\left\langle X_{i}\right\rangle_{i \in I}$ would have a greatest element, which is a contradiction. Thus, the index-set $I$ is infinite. Furthermore, as for every $i \in I$ there exists $x \in X$ such that $x \notin X_{i}$ the set $X$ is infinite as well.

In an algebraic cpo every element $x$ is the supremum of the finite elements below it. That is, if ( $D, \sqsubseteq$ ) is an algebraic cpo, then for all $x \in D$ we have

$$
x=\bigsqcup\{y \mid y \in D, y \sqsubseteq x, y \text { finite }\} .
$$

For example, the powerset cpo is an algebraic cpo as every (especially every infinite) set is the supremum (in this case the union) of its finite subsets. In an algebraic cpo the result of a (continuous) function for a non-finite element is the supremum of the results for the finite elements below. In other words, the behavior of functions for non-finite arguments is determined by the behavior for finite arguments. Note that

$$
\begin{aligned}
\llbracket \sigma \rightarrow \tau \rrbracket & =\{f \mid f \in \llbracket \sigma \rrbracket \rightarrow \llbracket \tau \rrbracket, f \text { continuous }\} \\
\llbracket \tau_{1} \times \cdots \times \tau_{n} \rrbracket & =\left\{\left\langle x_{1}, \ldots, x_{n}\right\rangle \mid x_{1} \in \llbracket \tau_{1} \rrbracket, \ldots, x_{n} \in \llbracket \tau_{n} \rrbracket\right\} \\
\llbracket \text { Bool } \rrbracket & =\{\perp, \text { False }, \text { True }\} \\
\llbracket \text { List } \tau \rrbracket & =\{\perp, \text { Nil }\} \cup\{\text { Cons }\langle x, x s\rangle \mid x \in \llbracket \tau \rrbracket, x s \in \llbracket \text { List } \tau \rrbracket\}
\end{aligned}
$$

Figure 2.2.2: Type Semantics
this characteristic of algebraic cpos is closely related to the behavior of functions in a functional programming language. In a functional programming language we cannot explicitly check whether the argument of a function is an infinite data structure, and, therefore, the behavior of this function for infinite arguments is determined by its behavior for finite arguments. This illustrates that we can use algebraic cpos to model the semantics of a functional programming language.

Figure 2.2.2 presents the interpretation of the types we are considering. The function $\llbracket \llbracket \rrbracket$ denotes type semantics as well as term semantics. We consider a standard type semantics, that is, we interpret functions as continuous mappings with a pointwise ordering. Here continuity means Scott-continuity, namely, monotonicity and preservation of suprema of chains. Functions are ordered point-wise, that means, we have $f \sqsubseteq g$ if and only if $f x \sqsubseteq g x$ for all $x$ of appropriate type.

As we consider a first-order language, all functions take a single argument. We denote mathematical function application by juxtaposition, that is, by terms of the form $f x$ where $f$ is a function and $x$ is an argument. The argument type of a syntactic function has the form $\tau_{1} \times \cdots \times \tau_{n}$ and is interpreted as non-lifted direct product. The elements of $\llbracket \tau_{1} \times \cdots \times \tau_{n} \rrbracket$ are denoted by $\left\langle e_{1}, \ldots, e_{n}\right\rangle$.

Because it is determined by the context, we use the symbol $\perp$ for all least elements and mostly do not distinguish least elements of different domains, for example, $\perp_{\text {Bool }}$ and $\perp_{\text {List Bool }}$. The type of Boolean values is interpreted by the three-element flat ordered set $\{\perp$,False, True $\}$ where $\perp$ is the least element. Note that we use the terms False and True for syntactic as well as for semantic objects.

The interpretation of List $\tau$ for some type $\tau$ contains terms over the constant $\mathrm{Nil}_{\tau}$ that represents an empty list of type $\tau$ and the two-ary constructor Cons $\langle\cdot, \cdot\rangle$ for non-empty lists. The first argument of Cons $\langle\cdot, \cdot\rangle$ is an element of the interpretation of $\tau$ while the second argument is an element of the interpretation of List $\tau$ again. These terms are ordered component-wise, that is, for elements $t$ and $t^{\prime}$ of $\llbracket L i s t ~ \tau \rrbracket$ we have $t \sqsubseteq t^{\prime}$ if we get $t^{\prime}$ by replacing subterms in $t$ that are $\perp$ by arbitrary values of appropriate type. Figure 2.2.3 illustrates the ordering of some of the elements of $\llbracket L i s t ~ \tau \rrbracket$, where $x$ and $y$ are elements of the interpretation of $\tau$.

The set of terms over Nil and Cons $\langle\cdot, \cdot\rangle$ and the interpretation of $\tau$ together with the component-wise ordering does not form a cpo. For example, consider the following infinite chain

$$
\perp, \text { Cons }\langle x, \perp\rangle, \text { Cons }\langle x, \text { Cons }\langle x, \perp\rangle\rangle, \text { Cons }\langle x, \text { Cons }\langle x, \text { Cons }\langle x, \perp\rangle\rangle\rangle, \ldots
$$



Figure 2.2.3: Ordering of Finite Lists
for some element $x$ of the interpretation of $\tau$. The supremum of this chain would be an infinite term. Therefore, we have to add representations for the infinite lists to $\llbracket L i s t \tau \rrbracket$. We consider the least fixpoint of the recursive definition of $\llbracket L i s t \tau \rrbracket$ that constitutes a cpo. Schmidt (1987) uses an inverse limit construction to show that such a least fixpoint exists. An element of $\llbracket$ List $\tau \rrbracket$ is represented by an infinite tuple of different approximations of the element itself. For example, the list Cons $\langle x$, Cons $\langle x, \perp\rangle\rangle$ is represented by the infinite tuple

$$
(\perp, \text { Cons }\langle x, \perp\rangle, \text { Cons }\langle x, \text { Cons }\langle x, \perp\rangle\rangle, \text { Cons }\langle x, \text { Cons }\langle x, \perp\rangle\rangle, \ldots)
$$

where the dots indicate that all following tuple components are Cons $\langle x$, Cons $\langle x, \perp\rangle\rangle$ as well. More precisely, we represent a finite list by an infinite tuple that becomes stationary at some position of the tuple. The components of the tuple contain approximations of the original lists that become more precise. That is, each consecutive pair of components of the tuple contains lists that are related by $\sqsubseteq$. The component where the tuple becomes stationary contains the original list. An infinite list is represented by an infinite tuple that contains approximations of the infinite list.

The inverse limit construction does not yield a solution to the recursive equation for the list type presented in Figure 2.2.2. However, we get a solution that is orderisomorphic to a solution of the concrete recursive definition. Therefore, in the following we consider the domain $\llbracket L i s t ~ \tau \rrbracket$ as the set of finite and infinite terms over Nil, Cons $\langle\cdot, \cdot\rangle$, and the interpretation of the content type $\tau$. Schmidt (1987) has presented a detailed construction of solutions to recursive domain equations.

Figure 2.2.4 presents the term semantics for the language shown in Figure 2.2.1. The semantics of a variable $x$ is its binding in the environment $a$. An environment is a mapping from the set of variables to semantic objects. By $a[x \mapsto v]$ we denote the modification of the environment $a$ such that $a[x \mapsto v](x)=v$ and $a[x \mapsto v](y)=$ $a(y)$ if $x \neq y$. Furthermore, we denote the modification of multiple variable bindings by $a\left[x_{1} \mapsto v_{1}, \ldots, x_{n} \mapsto v_{n}\right]$. If the semantics of an expression $e$ is independent of the environment $a$, in other words, $e$ has no free variables, then we use $\llbracket e \rrbracket$ instead of $\llbracket e \rrbracket_{a}$.

$$
\begin{aligned}
& \llbracket x \rrbracket_{a}=a(x) \\
& \llbracket \text { False } \rrbracket_{a}=\text { False } \\
& \llbracket \text { True } \rrbracket_{a}=\text { True } \\
& \llbracket N i l_{\tau} \rrbracket_{a}=\text { Nil } \\
& \llbracket \operatorname{Cons}\left\langle e_{1}, e_{2}\right\rangle \rrbracket_{a}=\text { Cons } \llbracket\left\langle e_{1}, e_{2}\right\rangle \rrbracket_{a} \\
& \llbracket f\left\langle e_{1}, \ldots, e_{n}\right\rangle \rrbracket_{a}=\llbracket f \rrbracket \llbracket\left\langle e_{1}, \ldots, e_{n}\right\rangle \rrbracket_{a} \\
& \llbracket\left\langle e_{1}, \ldots, e_{n}\right\rangle \rrbracket_{a}=\left\langle\llbracket e_{1} \rrbracket_{a}, \ldots, \llbracket e_{n} \rrbracket_{a}\right\rangle \\
& \llbracket f \rrbracket_{a}=\lambda\left\langle v_{1}, \ldots, v_{n}\right\rangle \cdot \llbracket e \rrbracket_{a\left[x_{1} \mapsto v_{1}, \ldots, x_{n} \mapsto v_{n}\right]} \\
& \text { if } f:: \tau_{1} \times \cdots \times \tau_{n} \rightarrow \tau ; f\left\langle x_{1}, \ldots, x_{n}\right\rangle=e \in P \\
& \llbracket \text { case } e \text { of }\left\{\text { False } \rightarrow e_{1} ; \text { True } \rightarrow e_{2}\right\} \rrbracket_{a}= \begin{cases}\perp & \text { if } \llbracket e \rrbracket_{a}=\perp \\
\llbracket e_{1} \rrbracket_{a} & \text { if } \llbracket e \rrbracket_{a}=\text { False } \\
\llbracket e_{2} \rrbracket_{a} & \text { if } \llbracket e \rrbracket_{a}=\text { True }\end{cases} \\
& \llbracket \text { case } e \text { of }\left\{N i l_{\tau} \rightarrow e_{1} ; \operatorname{Cons}\langle x, x s\rangle \rightarrow e_{2}\right\} \rrbracket_{a}= \begin{cases}\perp & \text { if } \llbracket e \rrbracket_{a}=\perp \\
\llbracket e_{1} \rrbracket_{a} & \\
\llbracket e_{2} \rrbracket_{a\left[x \mapsto v_{1}, x s \mapsto v_{2}\right]} \text { if } \llbracket e \rrbracket_{a}=N i \rrbracket_{\tau} & =\operatorname{Cons}\left\langle v_{1}, v_{2}\right\rangle\end{cases} \\
& \llbracket \text { undefined }_{\tau} \rrbracket_{a}=\perp_{\tau}
\end{aligned}
$$

Figure 2.2.4: Term Semantics

The constants False, True, and Nil are interpreted by the corresponding mathematical objects. The semantics of a list whose outermost constructor is Cons is the semantical function Cons $\langle\cdot, \cdot\rangle$ applied to the semantics of the arguments of the list constructor. The term $\lambda a . b$, used in the semantics of a function $f$, introduces an anonymous mathematical function. Furthermore, in this rule we use a globally defined program called $P$. Recursive function definitions result in recursive definitions of the semantic function $\llbracket \cdot \rrbracket$. We use the least solution of the resulting system of equations.
The semantics of a case expression performs a case distinction over the semantics of the scrutinee. Note that, if the semantics of the scrutinee has the form Cons $\langle x, x s\rangle$, then we have to adapt the environment $a$ accordingly.

Finally, the constant undefined ${ }_{\tau}$ is interpreted by the least element of the interpretation of $\tau$. This constant does not add expressiveness to the language as we could define undefined $_{\tau}$ as a non-terminating recursive function loop in the language itself.

To assign a type to a term we employ the typing rules presented in Figure 2.2.5. The typing rules are standard rules for a simply typed, monomorphic, functional language. A typing judgement has the form $\Gamma \vdash e:: \tau$ and states that an expression $e$ has the type $\tau$ under a type assumption - also called type environment - $\Gamma$. A type assumption $\Gamma$ has the form $\Gamma=\left\{x_{1}:: \tau_{1}, \ldots, x_{n}:: \tau_{n}\right\}$ and assigns types to the free variables of an expression. By $\Gamma, x:: \tau$ we denote splitting a type environment into the binding of a variable $x$ (here bound to a type $\tau$ ) and an environment $\Gamma$ that does not contain a binding for $x$.

$$
\begin{gathered}
\Gamma, x:: \tau \vdash x:: \tau \text { (VAR) } \quad \Gamma \vdash \text { False }:: \text { Bool (FALSE) } \\
\Gamma \vdash \text { True }:: \text { Bool (TRUE) } \quad \Gamma \vdash \text { Nil }_{\tau}:: \text { List } \tau \text { (NIL) } \\
\frac{\Gamma \vdash e_{1}:: \tau \quad \Gamma \vdash e_{2}:: \text { List } \tau}{\Gamma \vdash \operatorname{Cons}\left\langle e_{1}, e_{2}\right\rangle:: \text { List } \tau} \text { (CONS) } \\
\frac{\Gamma \vdash e_{1}:: \tau_{1} \quad \ldots \quad \Gamma \vdash e_{n}:: \tau_{n}}{\Gamma \vdash\left\langle e_{1}, \ldots, e_{n}\right\rangle:: \tau_{1} \times \cdots \times \tau_{n}} \text { (TUPLE) } \\
\frac{\Gamma \vdash f:: \sigma \rightarrow \tau \quad \Gamma \vdash e:: \sigma}{\Gamma \vdash f e:: \tau}\left(\text { APP) } \quad \frac{f:: \sigma \rightarrow \tau \in P}{\Gamma \vdash f:: \sigma \rightarrow \tau}\right. \text { (FUN) } \\
\frac{\Gamma \vdash e:: \text { Bool } \quad \Gamma \vdash e_{1}:: \tau \quad \Gamma \vdash e_{2}:: \tau}{\Gamma \vdash \operatorname{case} e \text { of }\left\{\text { False } \rightarrow e_{1} ; \text { True } \rightarrow e_{2}\right\}:: \tau}(\text { BCASE) } \\
\frac{\Gamma \vdash e:: \text { List } \tau \quad \Gamma \vdash e_{1}:: \tau^{\prime} \quad \Gamma, x:: \tau, x s:: \text { List } \tau \vdash e_{2}:: \tau^{\prime}}{\Gamma \vdash \operatorname{case} e \text { of }\left\{\text { Nil } \tau \rightarrow e_{1} ; \operatorname{Cons}\langle x, x s\rangle \rightarrow e_{2}\right\}:: \tau^{\prime}} \text { (LCASE) } \\
\Gamma \vdash \text { undefined } \tau:: \tau
\end{gathered}
$$

Figure 2.2.5: Typing Rules for the presented Functional Language

In the typing rule for a function we again use a globally defined program $P$. We call a program $P$ well-typed if every type signature that is annotated to a function is correct with respect to the typing rules of Figure 2.2.5. More precisely, a program $P$ is well-typed if we have $\varnothing \vdash f:: \sigma \rightarrow \tau$ for all functions $f:: \sigma \rightarrow \tau$ in the program $P$. In the following we assume that all programs $P$ are well-typed.

Example 2.2.1: As an example for the semantics, presented in Figure 2.2.4, we consider the function andL, which is frequently used as an example in the following. The function andL corresponds to the standard Boolean conjunction (\&\&) in Haskell and is defined as follows.

$$
\begin{aligned}
& \text { andL }:: \text { Bool } \times \text { Bool } \rightarrow \text { Bool } \\
& \text { andL }\langle x, y\rangle= \\
& \text { case } x \text { of } \\
& \text { False } \rightarrow \text { False } \\
& \text { True } \rightarrow y
\end{aligned}
$$

For any environment $a$ we calculate the semantics of andL as follows.

$$
\begin{aligned}
\llbracket a n d L \rrbracket_{a} & =\lambda\left\langle v_{1}, v_{2}\right\rangle . \llbracket \text { case } x \text { of }\{\text { False } \rightarrow \text { False; True } \rightarrow y\} \rrbracket_{a\left[x \mapsto v_{1}, y \mapsto v_{2}\right]} \\
& =\lambda\left\langle v_{1}, v_{2}\right\rangle . \begin{cases}\perp & \text { if } \llbracket x \rrbracket_{a\left[x \mapsto v_{1}, y \mapsto v_{2}\right]}=\perp \\
\llbracket \text { False } \rrbracket_{a\left[x \mapsto v_{1}, y \mapsto v_{2}\right]} & \text { if } \llbracket x \rrbracket_{a\left[x \mapsto v_{1}, y \mapsto v_{2}\right]}=\text { False } \\
\llbracket y \rrbracket_{a\left[x \mapsto v_{1}, y \mapsto v_{2}\right]} & \text { if } \llbracket x \rrbracket_{a\left[x \mapsto v_{1}, y \mapsto v_{2}\right]}=\text { True }\end{cases}
\end{aligned}
$$

$$
\begin{aligned}
& =\lambda\left\langle v_{1}, v_{2}\right\rangle \cdot \begin{cases}\perp & \text { if } v_{1}=\perp \\
\llbracket \text { False } \rrbracket_{a\left[x \mapsto v_{1}, y \mapsto v_{2}\right]} & \text { if } v_{1}=\text { False } \\
\llbracket y \rrbracket_{a\left[x \mapsto v_{1}, y \mapsto v_{2}\right]} & \text { if } v_{1}=\text { True }\end{cases} \\
& =\lambda\left\langle v_{1}, v_{2}\right\rangle \cdot \begin{cases}\perp & \text { if } v_{1}=\perp \\
\text { False } & \text { if } v_{1}=\text { False } \\
v_{2} & \text { if } v_{1}=\text { True }\end{cases}
\end{aligned}
$$

As the semantics of andL is independent of the environment $a$, we can use $\llbracket a n d L \rrbracket$ to refer to $\llbracket a n d L \rrbracket_{a}$.

2 Preliminaries

## 3 Non-Strict Evaluation

While we have already introduced the most important features of Haskell in Section 2.1, in this chapter we talk about its operational model. Haskell is a call-by need language, that is, programs are evaluated by so-called lazy evaluation (Wadsworth 1971; Launchbury 1993).

There are two main kinds of evaluation strategies, strict and non-strict evaluation. In a language with strict evaluation the arguments of a function are evaluated before the function is applied. In contrast, in a language with non-strict evaluation the arguments of a function are only evaluated if they are needed for the evaluation of the function application. Furthermore, there are two main kinds of non-strict evaluation strategies, call-by name and call-by need. In contrast to call-by name, the evaluation with call-by need avoids unnecessary re-evaluations, by sharing expressions. From a denotational point of view, these evaluation mechanisms are not distinguishable as they only differ with respect to efficiency. As we are mainly interested in the denotational aspects of evaluation, in the following, we only explain the call-by name aspects of non-strict evaluation in detail. Nevertheless, when we consider the effects of less strict implementations with respect to memory usage, as it is the case in Section 3.2 as well as in Chapter 7, we informally consider the effects of sharing.

In a language with call-by name evaluation, the arguments of a function are only evaluated if they are needed to determine the value of the application. For example, consider the following function, called const.

```
const : : \(\alpha \rightarrow \beta \rightarrow \alpha\)
const \(x_{-}=x\)
```

It takes a value of some type $\tau$ as first argument, ignores its second argument, and yields its first argument as result. When we evaluate an expression of the form const False expensive_computation, the expression expensive_computation is not evaluated as we can determine the value of the application without evaluating the second argument of const. More precisely, when const False expensive_computation is evaluated, const is replace by its right-hand side where the corresponding variables are substituted accordingly. Thus, the expression const False expensive_computation is replaced by False.

We can also use a run-time error or a non-terminating expression to observe that const does not cause the evaluation of its second argument. An expression that causes a run-time error or does not terminate when it is evaluated is denoted by $\perp$ in the following. For example, we have const False $\perp \equiv$ False where $\equiv$ denotes semantic equivalence. That is, even if we apply const False to a run-time error, this error is not raised because of non-strict evaluation. In the same way, even if the second argument of const is an expression whose evaluation does not terminate, the evaluation of the application of const terminates.

### 3.1 Advantages of Non-Strict Evaluation

From a practical point of view there are two main aspects of non-strict evaluation. First, we can make use of infinite data structures. Consider the following function, called iterate.

```
iterate :: (\alpha->\alpha)->\alpha->[\alpha]
iteratef }x=x\mathrm{ : iterate f (f x)
```

This function takes an initial value and constructs a list by repeated application of the provided function. Intuitively, we have iterate $f x \equiv[x, f x, f(f x), \ldots]$, where the dots denote that the list continues infinitely. More precisely, the evaluation of iterate $f x$ yields a list whose $n$-th element is generated by applying $f$ exactly $n-1$ times to the initial value $x$. For example, consider the expression iterate $(+1) 0$. When we evaluate this expression, we get an infinite list of increasing integers, starting with zero.

An infinite list seems to be of little interest, however in a non-strict language like Haskell we can make use of infinite data structures. For example, we can take an initial segment of an infinite list by means of the function take $::[\alpha] \rightarrow$ Int $\rightarrow[\alpha]$. This function takes a list and an integer and yields the initial segment of the list with the given number of elements. For example, when we evaluate the expression take 5 (iterate $(+1) 0$ ) we get the list $[0,1,2,3,4]$.

Infinite data structures allow for elegant definitions of certain algorithms. For example, we can define the Fibonacci function as follows. First, we define a constant fibs as the infinite list of Fibonacci numbers. The first two elements of the list of Fibonacci numbers are zero and one. We get the third element of the list by adding the first Fibonacci number to the second one, that is, we add the first element of fibs to the second element of fibs. We get the fourth element of the list by adding the second element to the third one. In general we simply use zipWith $(+)$ to point-wise add the elements of fibs to tail fibs where tail yields its argument list without the head.

$$
\begin{aligned}
& \text { fibs :: }[\text { Integer }] \\
& \text { fibs }=0: 1: \text { zipWith }(+) \text { fibs (tail fibs) }
\end{aligned}
$$

Note that we use the arbitrary precision integer type Integer for Fibonacci numbers as these numbers grow quite fast. By means of fibs we define a function fib that takes an integer and yields the corresponding Fibonacci number as follows.

$$
\begin{aligned}
& \text { fib }:: \text { Int } \rightarrow \text { Integer } \\
& \text { fib } i=\text { fibs }!!i
\end{aligned}
$$

By employing the list indexing function (!!), we simply project to a specific position and look up the corresponding Fibonacci number in the infinite list.

In the example of the Fibonacci numbers we can generate Fibonacci numbers without considering which number we are interested in. Thus, by using infinite data structures we can separate the generation from the consumption of data. Another
example for this style of programming is the calculation of the $n$-th prime number (Turner 1976). We define a function that generates the infinite list of prime numbers and get the $n$-th number by projecting to the $n$-th position of the list.

The other important practical aspect of non-strict evaluation has first been observed by Hughes (1989), namely, that non-strictness contributes to the modularity of programs. Haskell programs are often defined by means of predefined functions like map, filter, and zipWith. Hughes (1989) has observed that this style of programming heavily relies on non-strict evaluation.

For example, consider the function splitWhen $::(\alpha \rightarrow$ Bool $) \rightarrow[\alpha] \rightarrow[[\alpha]]$ from the Hackage ${ }^{1}$ package split by Yorgey (2010). The function splitWhen splits a list into a list of sublists. The list is split at all positions that satisfy the predicate, and, furthermore, the elements that satisfy the predicate are removed. For example, we have
splitWhen (== 'a') "abcada" 三 ["", "bc", "d", " " ].

Moreover, the predicate isAlpha :: Char $\rightarrow$ Bool from the standard library Data.Char yields True if and only if its argument is an alphabetic character. As a string is defined as a list of characters, we can use the application splitWhen (not $\circ$ isAlpha) to split a string into substrings that only contain alphabetical characters. For example, we have

$$
\text { splitWhen (not o isAlpha) "a. , bc d" } \equiv \text { ["a", "", "bc", "d"]. }
$$

Thus, by employing this function we can extract the words that occur in a text where the text is represented by a single string.

By means of splitWhen we can define the following function, called wordCount. It takes a word and a text and counts the number of occurrences of the word in the text.

```
wordCount :: String \(\rightarrow\) String \(\rightarrow\) Int
wordCount \(=\) length \(\circ\) filter \((==\) word \() \circ\) splitWhen ( \(n o t \circ\) isAlpha \()\)
```

First, we split the text into single words via splitWhen (not o isAlpha), then we remove all words that are not equal to the provided word by using filter, and, finally, we count the length of the filtered list.

Consider that we want to count occurrences of the string "mathematics" in a file that contains Shakespeare's collected works, which is freely available from Project Gutenberg. In a strict programming language the function splitWhen evaluates its argument completely. That is, the string that represents Shakespeare's collected works is placed in the memory. In contrast, in a non-strict programming language like Haskell the production of data by one function and the consumption of this data by another function can be interleaved. As soon as the data is consumed, the memory can be deallocated by the garbage collector. This way, functions in a modular programming style, which is very commonly used in Haskell, often have a small memory footprint (Hughes 1989). For example, Figure 3.1.1 shows the heap profile for counting occurrences of "mathematics" in Shakespeare's collected works, which has a size of about six megabyte. ${ }^{2}$

[^4]

Figure 3.1.1: Heap Profile of Counting Occurrences of "mathematics" in Shakespeare's Collected Works.

Sansom and Peyton Jones (1995) have presented a tool to generate heap profiles for Haskell programs. The current implementation of the Glasgow Haskell Compiler (GHC) provides an implementation of this kind of profiler. A heap profile plots the occupied heap memory against the time. The GHC provides different kinds of profiling. In this thesis we use the most basic form of profiling that distinguishes memory occupied by different data constructors and different types of objects like functions (FUN) and thunks (THUNK). We use this basic form of profiling as we do not investigate the memory usage by means of profiling but only use the profiles for illustrative purposes.

The profile of counting occurrences of "mathematics" shows that it needs less than three kilobyte of memory to process the complete file. The different shades of gray - or colors respectively - in the heap profile distinguish heap objects of different kinds. For example, the topmost - or yellow - part of the heap represents the amount of memory that is occupied by the stack of the program. The topmost but one - or pink - part of the heap profile represents the amount of memory that is occupied by occurrences of the list constructor (:) in the heap. The other parts of the profile are not of interest as these are mostly constant overheads caused by reading the content from a file and printing the result. In summary, we observe that the total amount of memory that is occupied at one time is quite small compared to the amount of data that is processed. And, even more impressively, the amount of memory that is occupied at one time stays the same even if we increase the size of the processed file.

### 3.2 Unnecessarily Strict Functions

Although Haskell is a non-strict programming language, functions can be unnecessarily strict. Consider the function intersperse from the standard library Data.List. It intersperses an element between all pairs of succeeding elements of a list.


Figure 3.2.1: Heap Profile of Replacing Umlauts in Shakespeare's Work with the Standard (left) and the Less Strict (right) Implementation of intersperse.

```
intersperse :: }\alpha->[\alpha]->[\alpha
intersperse _ [] = []
intersperse - [x] = [x]
intersperse sep (x:xs)=x: sep : intersperse sep xs
```

Furthermore, consider the function splitWhen $::(\alpha \rightarrow$ Bool $) \rightarrow[\alpha] \rightarrow[[\alpha]]$ from the Hackage package split again. By means of intersperse and splitWhen we define a function replaceBy, which replaces all occurrences of a specific element in a list by a given list.

$$
\begin{aligned}
& \text { replaceBy }:: \text { Eq } \alpha \Rightarrow \alpha \rightarrow[\alpha] \rightarrow[\alpha] \rightarrow[\alpha] \\
& \text { replaceBy } x \text { sep }=\text { concat } \circ \text { intersperse sep } \circ \text { splitWhen }(==x)
\end{aligned}
$$

First, by using splitWhen we split the argument list at all occurrences of the element $x$. Then, we intersperse the new separator sep between the elements of the resulting list, and, finally, we concatenate the resulting list via concat $::[[\alpha]] \rightarrow[\alpha]$. For example, we have replaceBy 'a' "bb" "mathematics" 三 "mbbthembbtics".

Let us consider an application of the function replaceBy. In HTML we have to replace all occurrences of German umlauts by a specific HTML encoding. For this example we only replace all occurrences of the character ' $\backslash 228$ ', which is the German umlaut "ä", by the corresponding encoding "\ä". Again, we process the collected works by Shakespeare. Note that we consider the english version of Shakespeare's collected works, and, therefore, there are no umlauts. We will comment on this issue at the end of this section. The left-hand side of Figure 3.2.1 presents the heap profile for escaping umlauts in Shakespeare's collected works, again, this profile it generated by the GHC profiler.

The behavior of replaceBy is quite contrary to the expectations about memory usage in a non-strict programming language. The program uses more than 50 megabyte of memory while we would have expected a profile similar to Figure 3.1.1. In Figure 3.2.1 the black part of the heap profile represents the amount of memory that
is occupied by the list constructor (:). The profile does not contain other parts because parts that sum to a total of less than $1 \%$ of the profile are removed from a profile.

Unexpected consumption of memory like this is called a space leak (Wadler 1987). In this case the space leak is caused by an unnecessarily strict implementation of intersperse. We can use Sloth, the tool that is presented in this thesis, to observe that the current implementation of intersperse is too strict. As Sloth is a light-weight tool, we only have to add the following line to our module to check whether a function is as little strict as possible.

## import Test.Sloth

This module provides a function called strictCheck to check whether a function is as little strict as possible for inputs up to a specific size. In the following we refer to a function that is as little strict as possible as minimally strict. In Chapter 4 we provide a formal definition of minimally strict functions. The size of a term is the number of constructors in the term. That is, strictCheck (intersperse :: $A \rightarrow[A] \rightarrow[A]) 4$ checks whether the $A$-instance of intersperse is as little strict as possible for inputs whose size in sum is at most four. Sloth generates test cases, that is, elements of the argument type of a function, to check whether a function is minimally strict. In the case of a polymorphic function this is hardly possible as we would have to check every possible monomorphic instance of the polymorphic function. Instead, we instantiate all type variables of a polymorphic function type with the opaque data type $A$, which is provided by the module Test.Sloth. In Chapter 6 we prove that a polymorphic function is indeed minimally strict if and only if its $A$-instance is minimally strict.

Sloth reports the following result if we check whether intersperse is minimally

```
> strictCheck (intersperse :: A -> [A] -> [A]) 4
3: \a (b:\perp) -> b:\perp
5: \a (b:c:\perp) -> b:a:c:\perp
Finished 7 tests.
```

It presents two argument result pairs, which show that intersperse is unnecessarily strict. The first one states that intersperse yields $\perp$ if it is applied to $a$ and $b: \perp$, where $a$ and $b$ are arbitrary values. A minimally strict implementation of intersperse yields $b: \perp$ for these arguments instead. ${ }^{3}$ Sloth highlights the subterm on the right-hand side of $->$ where the tested function is too strict. That is, it highlights a term that is not an error, while the current implementation yields an error at the highlighted position instead. Moreover, there exists a less strict implementation that yields the highlighted value. Note that Sloth does not only observe that intersperse is too strict for the arguments $a$ and $b: \perp$, but also that the first element of the result list has to be the first element of the argument list. This is possible due to the special treatment of polymorphic functions.

[^5]In the following we refer to an argument together with the current result of the function and the proposed result as a counter-example. We use the term counter-example as these informations show that the function is not minimally strict. The second counter-example for intersperse states that intersperse yields $b: a: \perp$ if we apply it to $a$ and $b: c: \perp$, while there exists a less strict implementation that yields $b: a: c: \perp$ for these arguments instead.

To provide a less strict implementation of intersperse we consider the following slightly different but equivalent implementation. We replace the pattern matching that checks whether the argument list has exactly one element by a case expression. Note that most compilers perform a similar transformation as many core languages use case expressions and rules that do not perform pattern matching. To transform intersperse into such a core language we would have to replace the pattern matching in the remaining rules by a case expression as well.

```
intersperse \(:: \alpha \rightarrow[\alpha] \rightarrow[\alpha]\)
intersperse _ [] \(=[]\)
intersperse sep \((x: x s)=\)
    case \(x s\) of
        [] \(\rightarrow x:\) []
        \({ }_{-} \rightarrow x\) : sep : intersperse sep \(x\) s
```

Using the information provided by Sloth we can define a less strict implementation of intersperse by yielding the first element of the list "before" performing pattern matching on the tail.

```
intersperse \(^{\prime}:: \alpha \rightarrow[\alpha] \rightarrow[\alpha]\)
intersperse \({ }^{\prime}\) - [] \(=[]\)
intersperse sep \((x: x s)=\)
    \(x\) : case \(x\) s of
        [] \(\rightarrow\) []
    \({ }_{-} \rightarrow\) sep:intersperse' sep xs
```

To get from intersperse to intersperse ${ }^{\prime}$ we distribute the partial application ( $x$ : ) over the case expression. We refer to this transformation as case deferment. In general, case deferment does not always yield a less strict implementation, but we always get an implementation that is at least as little strict as the original implementation. In Chapter 7 we consider this transformation in more detail and develop conditions when case deferment yields a less strict implementation.

There we observe that intersperse ${ }^{\prime}$ is indeed less strict than intersperse if there are arguments sep and $x: x s$ for which the evaluation of the scrutinee of the case expression yields an error. Furthermore, the context that is pulled over the case expression has to be more defined than an error if we apply the context to an error. More precisely, in this particular case, we have to provide arguments sep and $x: x s$ such that $x s$ is equivalent to an error and ( $x$ :) applied to an error is not an error. An arbitrary value sep of type $\tau$ and a list of the form $x: \perp$, where $x$ is an arbitrary value of type $\tau$ and $\perp$ denotes an error, satisfy these requirements. The results from Chapter 7 show that intersperse ${ }^{\prime}$ is, therefore, indeed less strict than intersperse.

The definition of intersperse ${ }^{\prime}$ still has a shortcoming, and, in fact, the definition of intersperse has the same shortcoming. The function intersperse ${ }^{\prime}$ performs the same pattern matching several times because intersperse ${ }^{\prime}$ checks whether $x$ s is the empty list and we perform the same check again in the recursive application. To calculate an improved implementation of intersperse we apply the worker/wrapper transformation by Gill and Hutton (2009). We do not present the calculation itself, but, for completeness, we provide the functions wrap and unwrap, used to derive the new implementation. We use the following functions.

```
wrap \(::(\alpha \rightarrow[\alpha] \rightarrow[\alpha]) \rightarrow \alpha \rightarrow[\alpha] \rightarrow[\alpha]\)
wrap \(_{--}[]=[]\)
wrap \(f \operatorname{sep}(x: x s)=x: f\) sep \(x \mathrm{~s}\)
unwrap \(::(\alpha \rightarrow[\alpha] \rightarrow[\alpha]) \rightarrow \alpha \rightarrow[\alpha] \rightarrow[\alpha]\)
unwrap \(_{--}[]=[]\)
unwrap \(f\) sep \(x s=\) sep : \(f\) sep \(x s\)
```

Using these functions for the worker/wrapper transformation we can derive the following function intersperse ${ }^{\prime \prime}$, which is equivalent to intersperse ${ }^{\prime}$ and, therefore, less strict than intersperse.

```
intersperse \({ }^{\prime \prime}:: \alpha \rightarrow[\alpha] \rightarrow[\alpha]\)
intersperse" \({ }^{\prime \prime}[]=[]\)
intersperse \({ }^{\prime \prime}\) sep \((x: x s)=x:\) work \(x\) s
    where
```

```
    work [] \(=\) []
```

    work [] \(=\) []
    work \((y: y s)=\) sep \(: y:\) work ys
    ```
    work \((y: y s)=\) sep \(: y:\) work ys
```

Because we only know that intersperse ${ }^{\prime \prime}$ is less strict than intersperse but not whether intersperse ${ }^{\prime \prime}$ is as little strict as possible, we can use Sloth to check whether intersperse ${ }^{\prime \prime}$ is still unnecessarily strict. Sloth does not report any counter-examples for inputs up to size 100.

The heap profile on the right-hand side of Figure 3.2 .1 shows the memory usage of escaping umlauts using the less strict implementation of intersperse. As both profiles in Figure 3.2.1 use the same scaling, the heap profile for the less strict implementation is barely notable. There is a small black line at the bottom of the graph which states that the process takes around 0.6 seconds. Figure 3.2.2 presents a magnified version of the heap profile of escaping umlauts using the less strict implementation intersperse ${ }^{\prime \prime}$. In this case the program uses less than three kilobyte of memory. That is, one delay of a pattern matching improves the memory usage by a factor of 20,000 .

Obviously, we seldomly generate HTML from a 6MB text file. Furthermore, we have chosen a file that does not contain any umlauts on purpose. The presented space leak is linear in the size of the longest substring that does not contain the character that is replaced. More precisely, intersperse does not yield a result until it is able to observe whether its argument has exactly one element. The function splitWhen checks whether any character of Shakespeare's collected works is an umlaut and when it finally arrives at the end the corresponding string yields a singleton list that contains Shakespeare's collected works. Therefore, splitWhen evaluates the string


Figure 3.2.2: Magnified Version of the Heap Profile of Replacing Umlauts in Shakespeare's Work with the Less Strict Implementation of intersperse.
that represents Shakespeare's collected works completely before it yields a result with one element. Furthermore, intersperse keeps the result of splitWhen in the memory as it does not yield the first element of its result until it has checked whether its argument has exactly one element. Therefore, we get a space usage that is linear in the size of Shakespeare's collected works.

In contrast to intersperse, as soon as splitWhen has observed that a character of Shakespeare's collected works is no umlaut, the function intersperse ${ }^{\prime \prime}$ passes this character on. Thus, replaceBy passes Shakespeare's collected works character-wise to its surrounding context if we use intersperse ${ }^{\prime \prime}$. Therefore, if we consider a well-behaved context, the garbage collector is able to deallocate parts of the string of Shakespeare's collected works immediately. Hence, this process never occupies a large amount of memory.

Although we have considered an input that does not contain umlauts, this is not an artificial example. The presented space leak is a modification of a space leak discovered by Fischer (2010) in his searchstring package (Fischer et al. 2010). Furthermore, a priori we often do not know whether a file contains a certain character or substring, and we also get a space leak if the character or substring occurs only rarely in the file. In summary, this example is supposed to emphasise that being too strict can have a tremendous effect with respect to memory behavior when processing large amounts of data.

3 Non-Strict Evaluation

## 4 Mathematical Model of Minimally Strict Functions

In this chapter we present the theoretical background for minimal strictness and the implementation of Sloth. In Section 4.1 we use the denotational semantics presented in Section 2.2 as the basis for the definition of least-strictness, which has been introduced by Chitil (2006) in an informal manner. We introduce the concept of sequentiality in Section 4.2 , which is used in Section 4.3 to define minimal strictness. Furthermore, we argue that the conecpt of minimally strict functions is an improvement over the concept of least strict functions. Finally, we provide a criterion to check whether a function is minimally strict and prove that this criterion is sufficient (Section 4.3.1) and necessary (Section 4.3.2). As mentioned in the introduction all proofs omited in this chapter are presented in Appendix A.

### 4.1 Least Strict Functions

Chitil (2006) has originally presented the idea of checking whether a function is as little strict as possible. In this section we present his approach and argue that it has a shortcoming that is undesirable in practice.

We start by defining a less-strict relation for functions. On basis of the semantics of a type we define the set of total values of a type. A total value is a maximal element of the interpretation of a type.

Definition 4.1.1 (Total Value): For a partially ordered set ( $P, \sqsubseteq$ ) we define the set of all maximal elements $P \uparrow \subseteq P$ as follows.

$$
P \uparrow:=\{x \in P \mid \forall y \in P . x \sqsubseteq y \Longrightarrow x=y\}
$$

In the following we refer to elements of $\llbracket \tau \rrbracket$ as values of type $\tau$, to elements of $\llbracket \tau \rrbracket \uparrow$ as total values of type $\tau$, and to elements of $\llbracket \tau \rrbracket \backslash \llbracket \tau \rrbracket \uparrow$ as partial values of type $\tau$. Furthermore, the name of a variable indicates its kind, more precisely, we use variable names that start with $v$ for values, $p v$ for partial values, and $t v$ for total values.

The next definition provides a formal treatment of the less-strict relation established by Chitil (2006). In other words, we define when we call a function $f$ less or equally strict than a function $g$.

Definition 4.1.2 (Less-Strict Relation): Let $f, g \in \llbracket \sigma \rightarrow \tau \rrbracket$. The functions $f$ and $g$ are related by $\preceq$ if the following holds.

$$
f \preceq g: \Longleftrightarrow \forall t v \in \llbracket \sigma \rrbracket \uparrow \cdot f t v=g t v \wedge \forall v \in \llbracket \tau \rrbracket \cdot f v \sqsupseteq g v
$$

If $f \prec g$, that is, $f \preceq g$ and $g \npreceq f$, we say that $f$ is less strict than $g$.
The relation $\preceq$ is reflexive, antisymmetric, and transitive because $=$ and $\sqsupseteq$ are reflexive, antisymmetric, and transitive; thus, $\preceq$ is a partial order.

Note that two functions are only related by $\preceq$ if they agree for total values. That is, we assume that it is intended if a function yields a non-total result for a total argument as we illustrate in the following example.

Example 4.1.1: Consider the function head that takes a list and yields its first element. As our simple functional language does not provide polymorphism, we consider the monomorphic Boolean instance of the head function. Naturally the head function yields a run-time error if it is applied to an empty list.

```
head :: List Bool \(\rightarrow\) Bool
head \(x s=\)
    case \(x\) of
        Nil \(_{\text {Bool }} \rightarrow\) undefined \(_{\text {Bool }}\)
        Cons \(\langle y, y s\rangle \rightarrow y\)
```

We get the following semantics for head.

$$
\llbracket h e a d \rrbracket=\lambda v \cdot \begin{cases}v_{1} & \text { if } v=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle \\ \perp & \text { otherwise }\end{cases}
$$

Now consider the function head' that yields a Boolean value, namely, False, in the case that its argument is the empty list.

```
head' \(::\) List Bool \(\rightarrow\) Bool
head' \(x s=\)
    case \(x s\) of
    Nil \(_{\text {Bool }} \rightarrow\) False
    Cons \(\langle y, y s\rangle \rightarrow y\)
```

We get the following semantics for head ${ }^{\prime}$.

$$
\llbracket h e a d^{\prime} \rrbracket=\lambda v . \begin{cases}\text { False } & \text { if } v=\operatorname{Nil}_{\tau} \\ v_{1} & \text { if } v=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle \\ \perp & \text { otherwise }\end{cases}
$$

If we omit the requirement that two functions $f$ and $g$ are only related by $\preceq$ if they agree for total values, then head' is less strict than head. Furthermore, if we replace False in the definition of head' by True we get another function that is less strict than head. Regarding a tool that checks whether a function is unnecessarily strict the tool would have to suggest an arbitrary total value as result for the empty list. However, more importantly, in the case of head the programmer intended to define a partial function. Therefore, we do not want to suggest a total function only for the sake of a less strict implementation.

Here and in the following we regard the behavior of a function for total values as the one a programmer cares about. That is, we might change the behavior of a function with respect to partial values but preserve its behavior with respect to total values. In summary, we want to check whether a function is as little strict as possible with respect to partial values while we preserve its behavior with respect to total values. The following example illustrates the less-strict relation by means of several implementations of the Boolean conjunction.

Example 4.1.2: Let us consider the Boolean conjunction andL from Example 2.2.1 and its strict counterpart and that is defined as follows.

$$
\begin{aligned}
& \text { and }:: \text { Bool } \times \text { Bool } \rightarrow \text { Bool } \\
& \text { and }\langle x, y\rangle= \\
& \text { case } x \text { of } \\
& \text { False } \rightarrow \text { case } y \text { of } \\
& \quad \begin{array}{r}
\text { False } \rightarrow \text { False } \\
\text { True } \rightarrow \text { False } \\
\text { True } \rightarrow \text { case } y \text { of } \\
\text { False } \rightarrow \text { False } \\
\text { True } \rightarrow \text { True }
\end{array}
\end{aligned}
$$

We get the following semantics for and.

$$
\llbracket a n d \rrbracket=\lambda\left\langle v_{1}, v_{2}\right\rangle \cdot \begin{cases}\perp & \text { if } v_{1}=\perp \vee v_{2}=\perp \\ \text { True } & \text { if } v_{1}=\text { True } \wedge v_{2}=\text { True } \\ \text { False } & \text { otherwise }\end{cases}
$$

For all total values of type Bool $\times$ Bool, $\langle$ False, False $\rangle,\langle$ False, True $\rangle,\langle$ True, False $\rangle$, and $\langle$ True, True〉, the function andL yields the same results as and. For the partial value $\langle$ False, $\perp\rangle$, the function andL yields False while and yields $\perp$. For all other partial values andL yields the same results as and. Therefore, by Definition 4.1.2 andL is less strict than and, that is, andL $\prec$ and. For readability, here and in the following, we sometimes identify a function $f$ with its semantics $\llbracket f \rrbracket$.

Let us consider the symmetric counterpart of andL, called andR. While andL performs pattern matching on its first argument, and $R$ performs pattern matching on its second argument.

$$
\begin{gathered}
\text { andR }:: \text { Bool } \times \text { Bool } \rightarrow \text { Bool } \\
\text { andR }\langle x, y\rangle= \\
\text { case } y \text { of } \\
\text { False } \rightarrow \text { False } \\
\text { True } \rightarrow x
\end{gathered}
$$

In comparison to the semantics of $\operatorname{andL}$, in the semantics of $a n d R$ the arguments $v_{1}$ and $v_{2}$ have changed their roles.

$$
\llbracket a n d R \rrbracket=\lambda\left\langle v_{1}, v_{2}\right\rangle \cdot \begin{cases}\perp & \text { if } v_{2}=\perp \\ \text { False } & \text { if } v_{2}=\text { False } \\ v_{1} & \text { if } v_{2}=\text { True }\end{cases}
$$

In the same manner as before, we can observe that andR is less strict than and. However, $a n d L$ and $a n d R$ are incomparable. While andL yields a more defined result than $\operatorname{andR}$ for the argument $\langle$ False, $\perp\rangle$, the function $\operatorname{andR}$ yields a more defined result than andL for the argument $\langle\perp$,False $\rangle$. Altogether we get andL $\prec$ and and andR $\prec$ and, but andR $\preceq$ andL and andL $\preceq$ andR.

So, how can we check whether a function is least strict or not? Let $f \in \llbracket \sigma \rightarrow \tau \rrbracket$. Because $f$ is monotonic, for every value $v$ of type $\sigma$ and every total value $t v$ of type $\sigma$ with $v \sqsubseteq t v$, we have $f v \sqsubseteq f t v$. Therefore, $f v$ is a lower bound of the set

$$
\{f t v \mid t v \in \llbracket \sigma \rrbracket \uparrow, v \sqsubseteq t v\} .
$$

As $f v$ is a lower bound of this set, it is less or equally defined than the corresponding greatest lower bound.

$$
f v \sqsubseteq \bigcap\{f t v \mid t v \in \llbracket \sigma \rrbracket \uparrow, v \sqsubseteq t v\}
$$

Chitil (2006) uses this inequality ${ }^{1}$ to check whether a function is least strict. A least strict function is supposed to agree with this greatest lower bound for all possible inputs $v$. The following definition states when a function is called least strict.

Definition 4.1.3 (Least-Strictness): For a function $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ we define a function $\inf _{f}: \llbracket \sigma \rrbracket \rightarrow \llbracket \tau \rrbracket$ as follows.

$$
\inf _{f} v=\Pi\{f t v \mid t v \in \llbracket \sigma \rrbracket \uparrow, v \sqsubseteq t v\}
$$

The function $f$ is called least strict if and only if $f=\inf _{f}$.
We have to prove that the infimum in the previous definition exists. First of all, all domains used here are interpreted as bounded complete cpos. A bounded complete cpo is also a complete meet-semilattice. In a complete meet-semilattice the infima of all non-empty subsets exist. That is, we have to show that the set

$$
\{f t v \mid t v \in \llbracket \sigma \rrbracket \uparrow, v \sqsubseteq t v\}
$$

is non-empty for all $v \in \llbracket \sigma \rrbracket$. We consider the following non-empty set of upper bounds of $v$.

$$
U:=\left\{v^{\prime} \in \llbracket \sigma \rrbracket \mid v \sqsubseteq v^{\prime}\right\}
$$

As $\llbracket \sigma \rrbracket$ is a cpo, every chain in $U$ has a least upper bound. By monotonicity this least upper bound is itself an element of $U$. Furthermore, $(U, \sqsubseteq)$ is a partially ordered set

[^6]and, by the Lemma of Zorn, $U$ has at least one maximal element. Let $m$ be a maximal element of $U$. We show that we have $m \in \llbracket \sigma \rrbracket \uparrow$ by employing the definition of $\llbracket \sigma \rrbracket \uparrow$. Let $y \in \llbracket \sigma \rrbracket$ such that $m \sqsubseteq y$. By transitivity we get $v \sqsubseteq y$ and, therefore, $y \in U$. As $m$ is maximal in $U$, we get $m=y$ and, thus, $m \in \llbracket \sigma \rrbracket \uparrow$. This shows that the set $\{f t v \mid t v \in \llbracket \sigma \rrbracket \uparrow, v \sqsubseteq t v\}$ is non-empty and, thus, has a greatest lower bound. In the following we frequently employ the fact that for every value $v$ of type $\sigma$ there exists a total value $t v$ of type $\sigma$ with $v \sqsubseteq t v$.

Note that $f$ and $\inf _{f}$ agree for total values, that is, we have $f t v=\inf _{f} t v$ for all total values $t v$. Furthermore, if a function $f$ is not least strict there exists a partial value $p v$ such that $f p v \sqsubset \inf _{f} p v$. Thus, if $f$ is not least strict the function inf ${ }_{f}$ is less strict than $f$. Hence, we are not only able to check whether a function is least strict but can also provide a less strict function, namely, $\inf _{f}$.
The definition of least-strictness has a shortcoming. Some functions that one would rank to be as little strict as possible are not least strict.

Example 4.1.3: Let us consider the function andL from Example 2.2.1 again. For the argument $\langle\perp$, False $\rangle$, the function andL yields $\perp$. On the other hand, we obtain the following equality.

$$
\begin{aligned}
\inf _{\llbracket a n d L \rrbracket}\langle\perp, \text { False }\rangle & =\rceil\{\llbracket a n d L \rrbracket t v \mid t v \in \llbracket \text { Bool } \times \text { Bool } \rrbracket \uparrow,\langle\perp, \text { False }\rangle \sqsubseteq t v\} \\
& =\rceil\{\llbracket a n d L \rrbracket\langle\text { False, False }\rangle, \llbracket \text { andL } \rrbracket\langle\text { True,False }\rangle\} \\
& =\rceil\{\text { False, False }\} \\
& =\text { False }
\end{aligned}
$$

That is, andL is not least strict because $\llbracket a n d L \rrbracket\langle\perp$, False $\rangle \neq \inf _{\llbracket a n d L \rrbracket}\langle\perp$, False $\rangle$.
Besides the definition of least-strictness, Chitil (2006) has presented a tool, which is called StrictCheck, that checks whether a Haskell function $f$ agrees with $\inf _{f}$ for all values up to a specific size. For example, the following application checks whether the Boolean conjunction ( $\& \&$ ) is least strict for values up to size ten. The function test2, provided by StrictCheck, takes an integer that specifies the maximal size of the generated test cases as first argument and a two-ary function as second argument.

```
Main> test2 10 (&&)
Function seems not to be least strict.
Input(s): ( }\perp,\mathrm{ False)
Current output: }
Proposed output: False
```

The result presented by test2 states that (\&\&) applied to an error (denoted by $\perp$ ) and False yields an error, but a least strict implementation is supposed to yield False instead. Note that (\&\&) is the Haskell function that corresponds to andL, which is defined in Example 2.2.1. That is, StrictCheck yields the result that we have observed in Example 4.1.3. Besides checking whether we have $\llbracket a n d L \rrbracket=\inf _{\llbracket a n d L \rrbracket}$ for all inputs, StrictCheck suggests

$$
\inf _{\llbracket a n d L \rrbracket}\langle\perp, \text { False }\rangle=\text { False }
$$

as result of $\llbracket a n d L \rrbracket$ for the argument $\langle\perp$,False $\rangle$ because we have

$$
\llbracket a n d L \rrbracket\langle\perp, \text { False }\rangle \neq \inf _{\llbracket a n d L \rrbracket}\langle\perp, \text { False }\rangle .
$$

Actually, andL is indeed not as little strict as possible. We can define a Boolean conjunction in Haskell that yields False for the argument $\langle$ False, $\perp\rangle$ as well as for the argument $\langle\perp$,False $\rangle$. For example, we can use the unamb operator, presented by Elliott (2009). The function unamb :: $\alpha \rightarrow \alpha \rightarrow \alpha$ takes two arguments and yields the argument that is not equal to $\perp$. If both arguments are $\perp$, it yields $\perp$ as well. To obtain this behavior even if $\perp$ is a non-terminating computation, unamb uses concurrency to fork two threads that evaluate the arguments of unamb in parallel. If one of the two threads is successful, unamb yields the result of this thread.

Because Haskell is a pure functional language, concurrency, as it is provided by Concurrent Haskell (Peyton Jones et al. 1996), for example, is performed in the IO monad (Peyton Jones and Wadler 1993). Thus, to define a function of type $\alpha \rightarrow \alpha \rightarrow$ $\alpha$ that uses concurrency we have to employ the function unsafePerformIO :: IO $\alpha \rightarrow \alpha$, which takes an I/O action and yields a non-monadic value. As its name suggests, unsafePerformIO is unsafe in the sense that it might destroy referential transparency. For example, the expression unamb False True might be evaluated to False or True, depending on the scheduling of the forked threads. Therefore, unamb breaks referential transparency as we cannot replace equals for equals. For example, the evaluation of let $x=$ unamb False True in $(x, x)$ might yield another result than the evaluation of (unamb False True, unamb False True) as the two occurrences of unamb False True might yield different results. Nevertheless, the function unamb is referential transparent if its arguments have a common least upper bound.

In practice we often want to know whether there is a less strict implementation that avoids the use of features like concurrency and unsafePerformIO. In fact, many functions are not least strict because there is often a bias towards one argument with respect to strictness. By using StrictCheck we have observed, that the functions (\&\&), (\|), (+), and, or, zip, as well as the list instances of ( $==$ ) and compare are not least strict, for example.

As an example, we examine the monomorphic instance of the equality check ( $==$ ) for lists of Boolean values. StrictCheck yields the following counter-example if we check $(==)::[$ Bool $] \rightarrow[$ Bool $] \rightarrow$ Bool for Boolean lists up to size 10.

```
Main> test2 10 ((==) :: [Bool] -> [Bool] -> Bool)
Function seems not to be least strict.
Input(s): ([False], [ }\perp, False]
Current output: \perp
Proposed output: False
```

That is, we are supposed to define an equality check that yields False for the arguments False : [] and $\perp$ : False : []. Furthermore, as the current implementation of (==) yields False for the arguments False : $\perp$ and True : $\perp$, a less strict implementation of the equality check is supposed to yield False as well. However, we cannot define a less strict implementation that satisfies these requirements and does not use features like concurrency, and it is not necessarily easy to observe this fact.

In contrast to StrictCheck, Sloth, the tool presented in this thesis, identifies (\&\&), $(\|),(+)$, and, or, zip, as well as the list instances of $(==)$ and compare as minimally strict as it checks whether a function is too strict without allowing features like concurrency. To restrict the considered functions we employ the concept of sequentiality by Vuillemin (1974). In the following section we introduce sequentiality and the concept of minimally strict functions, which is based on sequentiality.

### 4.2 Sequential and Demanded Positions

In order to restrict the considered first-order functional language to functions that do not use features like concurrency, we employ the definition of sequentiality by Vuillemin (1974). The definition of sequentiality employs contexts, which we only informally introduce here. Let $\Sigma$ be the set of constructor and function symbols. A context is a term over $\Sigma$ and the additional symbol [ $\cdot]$, called hole. Let $C$ be a context with $n$ holes, that is, $n$ occurrences of the symbol $[\cdot]$, and $e_{1}, \ldots, e_{n}$ be expressions. Then, $C\left[e_{1}, \ldots, e_{n}\right]$ denotes the context $C$ where the $i$-th hole is filled in with expression $e_{i}$. Vuillemin (1974) has given the following definition of a sequential language.
Definition 4.2.1 (Sequentiality): A functional language is called sequential if it satisfies the following. Let $C$ be a context with $n$ holes and

$$
\llbracket C\left[\text { undefined }_{\tau_{1}}, \ldots, \text { undefined }_{\tau_{n}}\right] \rrbracket=\perp .
$$

Then, there exists at least one $i \in\{1, \ldots, n\}$ such that

$$
\llbracket C\left[e_{1}, \ldots, e_{i-1}, \text { undefined }_{\tau_{i}}, e_{i+1}, \ldots, e_{n}\right] \rrbracket=\perp
$$

for all expressions $e_{1}, \ldots, e_{n}$ of appropriate types.
The following example illustrates the definition of sequentiality.
Example 4.2.1: By using the definition of sequentiality we can show that a Boolean conjunction that is less strict than andL is not sequential. We consider a Boolean conjunction andL' and a context $C$ defined as and $L^{\prime}\langle[\cdot],[\cdot]\rangle$. We have

$$
\begin{aligned}
\llbracket C\left[\text { undefined }_{\text {Bool }}, \text { undefined }_{\text {Bool }}\right\rfloor \rrbracket & =\llbracket \text { andL }^{\prime}\left\langle\text { undefined }_{\text {Bool }}, \text { undefined }_{\text {Bool }}\right\rangle \rrbracket \\
& =\llbracket \text { andL }^{\prime} \rrbracket \llbracket\left\langle\text { undefined }_{\text {Bool }} \text {,undefined }_{\text {Bool }}\right\rangle \rrbracket \\
& =\llbracket \text { andL }^{\prime} \rrbracket\langle\perp, \perp\rangle \\
& =\perp,
\end{aligned}
$$

that is, we can apply Definition 4.2.1 to the context $C$. The last step of this equation follows from monotonicity of $\llbracket a n d L^{\prime} \rrbracket$ and the specification of a Boolean conjunction, more precisely, $\llbracket a n d L^{\prime} \rrbracket\langle$ True, False $\rangle=$ False and $\llbracket a n d L^{\prime} \rrbracket\langle$ True, True $\rangle=$ True.

In a sequential language, and $L$ has to be strict with respect to one of its arguments. More precisely, we have

$$
\llbracket C\left[\text { undefined }_{\text {Bool }}, b\right] \rrbracket=\llbracket \text { andL }^{\prime}\left\langle\text { undefined }_{\text {Bool }}, b\right\rangle \rrbracket=\llbracket \text { andL } L^{\prime} \rrbracket\langle\perp, \llbracket b \rrbracket\rangle=\perp
$$

for all Boolean expressions $b$ or we have

$$
\llbracket C\left[b, \text { undefined }_{\text {Bool }}\right\rfloor \rrbracket=\llbracket \text { andL }^{\prime}\left\langle b, \text { undefined }_{\text {Bool }}\right\rangle \rrbracket=\llbracket \text { andL }^{\prime} \rrbracket\langle\llbracket b \rrbracket, \perp\rangle=\perp
$$

for all Boolean expressions $b$. Because $a n d L^{\prime}$ is less strict than andL, we have

$$
\llbracket a n d L^{\prime} \rrbracket\langle\text { False }, \perp\rangle \sqsupseteq \llbracket \text { andL } \rrbracket\langle\text { False }, \perp\rangle=\text { False } .
$$

This property together with sequentiality implies

$$
\llbracket a n d L^{\prime} \rrbracket\langle\perp, \text { False }\rangle=\perp .
$$

As we have observed in Example 4.1.3, a least strict Boolean conjunction is supposed to yield False for the argument $\langle\perp$, False $\rangle$. However, this application of sequentiality shows that there is no least strict Boolean conjunction in a sequential language.

To adapt the definition of sequentiality to our needs we introduce the concepts of positions, projections, and substitutions. The projections presented here are a generalization of the continuous projections that come along with the direct product. We start with the definition of the set of positions.

Definition 4.2.2 (Position): A position is a sequence of natural numbers. We denote the set of positions by $\mathbb{N}^{*}$, that is, the free monoid over the natural numbers. The empty sequence is denoted by $\epsilon$. If $n$ is a natural number and $p$ is a position, then $n . p$ is a position. Instead of $n . \epsilon$ we synonymously use $n$. For positions $p$ and $p^{\prime}$ we define that $p \leq p^{\prime}$ if and only if $p$ is a prefix of $p^{\prime}$.

Before we define projections and substitutions, we define the set of valid positions of an element of $\llbracket \tau \rrbracket$ for some type $\tau$. For simplicity, we define a single function that takes an element of the interpretation of an arbitrary type and yields the set of valid positions of this element. That is, we define a function Pos $: \bigcup_{\tau \in \text { Types }} \llbracket \tau \rrbracket \rightarrow 2^{\mathbb{N}^{*}}$ where $\mathbb{N}^{*}$ denotes the set of positions. As we want to define this function recursively, we define it by means of a least fixpoint. Therefore, it is advantageous if domain and range of Pos are cpos. As mentioned before, the powerset of an arbitrary set constitutes a cpo. As the domain of Pos is not a cpo, we add a least element denoted by $\perp$ to the set $\bigcup_{\tau \in \text { Types }} \llbracket \tau \rrbracket$. More precisely, we have $\perp \sqsubset \perp_{\tau}$ for all $\tau \in$ Types. Furthermore, two elements from the same cpo $\llbracket \tau \rrbracket$ are related by $\sqsubseteq$ if they are related by $\sqsubseteq$ in $\llbracket \tau \rrbracket$, and two elements from different cpos are incomparable. This construction is sometimes called separated sum as we sum up several cpos but separate the sets in the sense that we distinguish least elements of different cpos. Figure 4.2.1 illustrates the structure of the resulting cpo. In the following we abbreviate the set $\{\perp\} \cup \bigcup_{\tau \in \text { Types }} \llbracket \tau \rrbracket$ to $\Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$.

Definition 4.2.3 (Valid Position): We define a function Pos : $\Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \rightarrow 2^{\mathbb{N}^{*}}$ that yields the set of valid positions for an element of the interpretation of some type $\tau$. By Types we denote the set of first-order types, that is, the set Types contains tuples of arbitrary arity, Booleans, and list types. As the domain of Pos contains an additional least element $\perp$, we define Pos $\perp=\varnothing$.


Figure 4.2.1: Separated Sum of some Cpos $\llbracket \tau_{1} \rrbracket$ to $\llbracket \tau_{n} \rrbracket$
For the elements of the interpretation of a tuple type $\tau_{1} \times \cdots \times \tau_{n}$ we define the set of valid positions as follows.

$$
\operatorname{Pos}\left\langle v_{1}, \ldots, v_{n}\right\rangle=\bigcup_{i \in\{1, \ldots, n\}}\left\{i . p \mid p \in \operatorname{Pos} v_{i}\right\}
$$

In other words, we precede each valid position of one of the components of the tuple with the position of the component in the tuple.

For the elements of the interpretation of the type Bool we define the set of valid positions as follows.

$$
\text { Pos } \perp_{\text {Bool }}=\{\epsilon\} \quad \text { Pos False }=\{\epsilon\} \quad \text { Pos True }=\{\epsilon\}
$$

That is, the only valid position of any Boolean value is the root position $\epsilon$.
Next, we define Pos for the elements of the interpretation of List $\tau$. Let $v_{1} \in \llbracket \tau \rrbracket$ and $v_{2} \in \llbracket L i s t \tau \rrbracket$. We define the set of valid positions of a list as follows.

$$
\begin{gathered}
\operatorname{Pos} \perp_{\text {List } \tau}=\{\epsilon\} \quad \operatorname{Pos} \operatorname{Nil_{\tau }}=\{\epsilon\} \\
\operatorname{Pos}\left(\operatorname{Cons}\left\langle v_{1}, v_{2}\right\rangle\right)=\{\epsilon\} \cup\left\{1 . p \mid p \in \operatorname{Pos} v_{1}\right\} \cup\left\{2 . p \mid p \in \operatorname{Pos} v_{2}\right\}
\end{gathered}
$$

To solve this recursive definition we use the least solution of this equation. As the set $\Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$ as well as $2^{\mathbb{N}^{*}}$ constitute cpos and the corresponding functional is monotonic, the least fixpoint of this functional exists.

We give some examples of valid positions in the following.
Example 4.2.2: We consider the Boolean list Cons $\left\langle\perp_{\text {Bool }}, N i l_{\text {List Bool }}\right\rangle$. We have

$$
\operatorname{Pos}\left(\text { Cons }\left\langle\perp_{\text {Bool }}, N i l_{\text {List Bool }}\right\rangle\right)=\{\epsilon, 1,2\},
$$

that is, the root position as well as the first and the second argument of Cons are valid positions.

If we consider the tuple $\left\langle\right.$ False, Cons $\left.\left\langle\perp_{\text {Bool }}, N i l_{\text {List Bool }}\right\rangle\right\rangle$, which contains False in the first component and the Boolean list considered above in the second component, we have

$$
\operatorname{Pos}\left(\left\langle\text { False, Cons }\left\langle\perp_{\text {Bool }}, \text { Nil } L_{\text {List Bool }}\right\rangle\right\rangle\right)=\{1,2,2.1,2.2\} .
$$

That is, the first and the second component of the tuple are valid positions and the two arguments of Cons are valid positions.

Next, we define a projection function of type $\Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \times \mathbb{N}^{*} \rightarrow \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$ that takes a term $e$ and a position $p$ and yields the sub-term at that particular position denoted by $\left.e\right|_{p}$. Intuitively, we want to project to valid positions only. Thus, we could simply define a partial function that is not defined if the position is not a valid position of the argument of the projection. However, consider a value like the list Cons $\left\langle\perp_{\text {Bool }}, \perp_{\text {List Bool }}\right\rangle$. If we project to position 1 of this list, we want to get the first element of the list, that is, we would like to have

$$
\left.\left(\text { Cons }\left\langle\perp_{\text {Bool }}, \perp_{\text {List Bool }}\right\rangle\right)\right|_{1}=\perp_{\text {Bool }} \text {. }
$$

For proving statements involving projections it is advantageous if the projection function is continuous. Now let us consider the supremum

$$
\perp_{\text {List Bool }} \sqcup \text { Cons }\left\langle\perp_{\text {Bool }}, \perp_{\text {List Bool }}\right\rangle \text {. }
$$

By continuity we would get

$$
\begin{aligned}
\left.\left(\text { Cons }\left\langle\perp_{\text {Bool, }} \perp_{\text {List Bool }}\right\rangle\right)\right|_{1} & =\left.\left(\perp_{\text {List Bool }} \sqcup \text { Cons }\left\langle\perp_{\text {Bool }}, \perp_{\text {List Bool }}\right\rangle\right)\right|_{1} \\
& =\left.\left.\perp_{\text {List Bool }}\right|_{1} \sqcup\left(\text { Cons }\left\langle\perp_{\text {Bool }}, \perp_{\text {List Bool }}\right\rangle\right)\right|_{1} .
\end{aligned}
$$

If $\cdot \mid$. is a partial function, then $\left.\perp_{\text {List Bool }}\right|_{1}$ is not defined as we have Pos $\perp_{\text {List Bool }}=$ $\{\epsilon\}$. Therefore, instead of defining a partial projection function we define a function that yields the least element $\perp$ if we project to a non-valid position.

Definition 4.2.4 (Projection): We define a projection

$$
\cdot \mid: \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \times \mathbb{N}^{*} \rightarrow \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket
$$

that takes a term and a position and projects to the corresponding sub-term.
For all $\left\langle v_{1}, \ldots, v_{n}\right\rangle \in \llbracket \tau_{1} \times \cdots \times \tau_{n} \rrbracket$ and all valid positions i.p of $\left\langle v_{1}, \ldots v_{n}\right\rangle$ we define the projection as follows.

$$
\left.\left\langle v_{1}, \ldots, v_{n}\right\rangle\right|_{i . p}=\left.v_{i}\right|_{p}
$$

Note that we do not have to define the projection to the root position as $\epsilon$ is not a valid position of a tuple, in other words, we have $\epsilon \notin \operatorname{Pos}\left\langle v_{1}, \ldots, v_{n}\right\rangle$.

For the elements of the interpretation of the type Bool we define a projection to a valid position as follows.

$$
\left.\perp_{\text {Bool }}\right|_{\epsilon}=\perp_{\text {Bool }} \quad \text { False }\left.\right|_{\epsilon}=\text { False } \quad \text { True }\left.\right|_{\epsilon}=\text { True }
$$

Let $v_{1} \in \llbracket$ List $\tau \rrbracket, v_{2} \in \llbracket \tau \rrbracket$, and $p \in \operatorname{Pos}\left(\operatorname{Cons}\left\langle v_{1}, v_{2}\right\rangle\right)$. We define the projection to a valid position of a list as follows.

$$
\begin{aligned}
& \left.\perp_{\text {List } \tau}\right|_{\epsilon}=\perp_{\text {List } \tau} \\
& \left.\quad N i l_{\tau}\right|_{\epsilon}=N i l_{\tau} \\
& \left.\left(\text { Cons }\left\langle v_{1}, v_{2}\right\rangle\right)\right|_{p}= \begin{cases}\text { Cons }\left\langle v_{1}, v_{2}\right\rangle & \text { if } p=\epsilon \\
\left.v_{1}\right|_{q} & \text { if } p=1 . q \\
\left.v_{2}\right|_{q} & \text { if } p=2 . q\end{cases}
\end{aligned}
$$

If we project to a non-valid position, the projection yields $\perp$. That is, for all $v \in$ $\Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$ and $p \in \mathbb{N}^{*} \backslash$ (Posv) we define $\left.v\right|_{p}=\perp$. We, again, use the least solution of the recursive definition of $\cdot \mid$..

We give some examples of projections in the following.
Example 4.2.3: We consider the value $\left\langle\right.$ False, Cons $\left.\left\langle\perp_{\text {Bool }}, N i l_{\text {List Bool }}\right\rangle\right\rangle$ and abbreviate it to $v$ in the following. Example 4.2.2 presents the set of valid positions for this value. We have

$$
\left.v\right|_{1}=\text { False },\left.\quad v\right|_{2}=\text { Cons }\left\langle\perp_{\text {Bool }}, N i l_{\text {List Bool }}\right\rangle, \quad \text { and }\left.\quad v\right|_{2.2}=\operatorname{Nil}_{\text {List Bool }} .
$$

Furthermore, we have $\left.v\right|_{\epsilon}=\perp,\left.v\right|_{2.3}=\perp$, and $\left.v\right|_{2.1 .1}=\perp$ because $\epsilon, 2.3$, and 2.1.1 are not valid positions of $v$.

In the same manner as we have defined the projection function we define a substitution of type $\Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \times \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \times \mathbb{N}^{*} \rightarrow \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$ that takes two terms $e_{1}$ and $e_{2}$ and a position $p$ and substitutes a sub-term of the first term by the second term denoted by $e_{1}\left[e_{2}\right]_{p}$.

Definition 4.2.5 (Substitution): We define a substitution

$$
\cdot[\cdot] .: \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \times \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \times \mathbb{N}^{*} \rightarrow \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket .
$$

Let $\left\langle v_{1}, \ldots, v_{n}\right\rangle \in \llbracket \tau_{1} \times \cdots \times \tau_{n} \rrbracket$ and i.p be a valid position of $\left\langle v_{1}, \ldots, v_{n}\right\rangle$. Because i.p is a valid position, there exists $\tau^{\prime}$ such that $\left.\left\langle v_{1}, \ldots, v_{n}\right\rangle\right|_{\text {i.p }} \in \llbracket \tau^{\prime} \rrbracket$. For all $v \in \llbracket \tau^{\prime} \rrbracket$ we define the substitution of a valid position of a tuple as follows.

$$
\left\langle v_{1}, \ldots, v_{n}\right\rangle[v]_{i . p}=v_{i}[v]_{p}
$$

For all values $v$ of $\llbracket B \operatorname{Bool} \rrbracket$ we define the substitution of a valid position of an element of the interpretation of Bool as follows.

$$
\perp_{\text {Bool }}[v]_{\epsilon}=v \quad \text { False }[v]_{\epsilon}=v \quad \text { True }[v]_{\epsilon}=v
$$

Let $v \in \llbracket L i s t \tau \rrbracket$. We define the substitution of a list as follows.

$$
\perp_{\text {List } \tau}[v]_{\epsilon}=v \quad \operatorname{Nil}_{\tau}[v]_{\epsilon}=v
$$

Furthermore, let $v_{1} \in \llbracket \tau \rrbracket, v_{2} \in \llbracket L i s t ~ \tau \rrbracket$, and $p \in \operatorname{Pos}\left(\operatorname{Cons}\left\langle v_{1}, v_{2}\right\rangle\right)$. Because $p$ is a valid position, there exists $\tau^{\prime}$ such that $\left.\left(\right.$ Cons $\left.\left\langle v_{1}, v_{2}\right\rangle\right)\right|_{p} \in \llbracket \tau^{\prime} \rrbracket$. For all $v \in \llbracket \tau^{\prime} \rrbracket$ we define the substitution of a valid position of a list by $v$ as follows.

$$
\left(\text { Cons }\left\langle v_{1}, v_{2}\right\rangle\right)[v]_{p}= \begin{cases}v & \text { if } p=\epsilon \\ \text { Cons }\left\langle v_{1}[v]_{q}, v_{2}\right\rangle & \text { if } p=1 . q \\ \text { Cons }\left\langle v_{1}, v_{2}[v]_{q}\right\rangle & \text { if } p=2 . q\end{cases}
$$

If we substitute a non-valid position or the second argument has the wrong type, the substitution yields $\perp$. More precisely, for all $v, v^{\prime} \in \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$, and $p \in \mathbb{N}$ we define $v\left[v^{\prime}\right]_{p}=\perp$ if $p \notin \operatorname{Pos} v$. Furthermore, if $\left.v\right|_{p} \in \llbracket \tau^{\prime} \rrbracket$ and $v^{\prime} \notin \llbracket \tau^{\prime} \rrbracket$ we define $v\left[v^{\prime}\right]_{p}=\perp$. To solve the recursive definition we, again, use its least solution.

We give some examples of substitutions in the following.
 it to $v$ in the following. Example 4.2.2 presents to set of valid positions for this value. We have

$$
\begin{aligned}
v[\text { True }]_{1} & =\left\langle\text { True, Cons }\left\langle\perp_{\text {Bool }}, \text { Nil }_{\text {List Bool }}\right\rangle\right\rangle, \\
v\left[\perp_{\text {List Bool }}\right]_{2} & =\left\langle\text { False, } \perp_{\text {List Bool }}\right\rangle,
\end{aligned}
$$

and

$$
v\left[\text { Cons }\left\langle\text { False }, \perp_{\text {List Bool }}\right\rangle\right]_{2.2}=\left\langle\text { False, Cons }\left\langle\perp_{\text {Bool }} \text {, Cons }\left\langle\text { False, } \perp_{\text {List Bool }}\right\rangle\right\rangle\right\rangle .
$$

For all $v^{\prime} \in \Sigma_{\tau \in T y p e s} \llbracket \tau \rrbracket$ we have $v\left[v^{\prime}\right]_{\epsilon}=\perp, v\left[v^{\prime}\right]_{2.3}=\perp$, and $v\left[v^{\prime}\right]_{2.1 .1}=\perp$ because $\epsilon, 2.3$, and 2.1.1 are not valid positions of $v$. Furthermore, we have $v\left[\mathrm{Nil}_{\text {Bool }}\right]_{1}=\perp$ because $\left.v\right|_{1}=$ False $\in \llbracket$ Bool $\rrbracket$ and $N i l_{\text {Bool }} \in \llbracket L i s t$ Bool $\rrbracket$.

The following lemma states some properties about projections and substitutions that are employed in several of the following proofs.

Lemma 4.2.1: Projections and substitutions satisfy the following properties.

- For all positions $p$ the function $\left.\cdot\right|_{p}: \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \rightarrow \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$ is continuous.
- For all positions $p$ the function $\cdot[\cdot]_{p}: \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \times \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \rightarrow \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$ is continuous.
- For all positions $p$ and all $v \in \llbracket \tau \rrbracket$ the function $v[\cdot]_{p}: \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket \rightarrow \Sigma_{\tau \in \text { Types }} \llbracket \tau \rrbracket$ is strictly increasing if the results differ from $\perp$.

Note that continuity of $\cdot[\cdot]_{p}$ means that for all chains $\left\langle v_{i}\right\rangle_{i \in I}$ and $\left\langle v_{i}^{\prime}\right\rangle_{i \in I}$ we have

$$
\left(\bigsqcup_{i \in I} v_{i}\right)\left[\bigsqcup_{i \in I} v_{i}^{\prime}\right]_{p}=\bigsqcup_{i \in I}\left(v_{i}\left[v_{i}^{\prime}\right]_{p}\right) .
$$

Furthermore, note that continuity of $\left.\cdot\right|_{p}$ implies monotonicity of $\left.\cdot\right|_{p}$ and continuity of $\cdot[\cdot]_{p}$ implies monotonicity of $\cdot[v]_{p}$ for a fixed $v \in \llbracket \tau \rrbracket$. Finally, note that the restriction


Figure 4.2.2: Graphics illustrating $\left.v^{\prime}\right|_{p}$ and $v\left[\left.v^{\prime}\right|_{p}\right]_{p}$


Figure 4.2.3: Graphics illustrating $\left.\left(v\left[\left.v^{\prime}\right|_{p}\right]_{p}\right)\right|_{p^{\prime}}=\left.v^{\prime}\right|_{p^{\prime}}$ if $p^{\prime} \geq p$
to non- $\perp$ results in the last property in fact restricts the property to valid arguments with respect to position and type. To extend the presented approach to other data types the corresponding projections and substitutions have to satisfy the properties presented in Lemma 4.2.1 as well.

Besides these semantics related properties, there are some properties that specify a connection between projections and substitutions. The first property states that projections and substitutions are a kind of inverse to each other in the sense that we have $v\left[\left.v\right|_{p}\right]_{p}=v$ for all values of some interpretation $\llbracket \tau \rrbracket$ and valid positions $p$. The second property states that $p^{\prime} \geq p$ implies $\left.\left(v\left[\left.v^{\prime}\right|_{p}\right]_{p}\right)\right|_{p^{\prime}}=\left.v^{\prime}\right|_{p^{\prime}}$. We informally illustrate this equation in the following.

Consider the graphic on the left-hand side of Figure 4.2.2. The white triangle represents the term $v^{\prime}$, and the gray triangle represents the sub-term at position $p$, that is, $\left.v^{\prime}\right|_{p}$. In the graphic on the right-hand side the white triangle represents another term $v$. We replace the sub-term of $v$ at position $p$ by $\left.v^{\prime}\right|_{p}$. That is, at position $p$ we insert the gray triangle into the white triangle. Now consider that we project to a position $p^{\prime}$ with $p^{\prime} \geq p$ in the term $v\left[\left.v^{\prime}\right|_{p}\right]_{p}$. On the left-hand side of Figure 4.2 .3 this sub-term is represented by the dark gray triangle. As we have $p \leq p^{\prime}$ ( $p$ is a prefix of $p^{\prime}$ ), we project to a term that is a sub-term of the light gray triangle. Thus, we get the same result if we project to position $p^{\prime}$ of the original term where the light gray triangle is taken from. In other words we have $\left.\left(v\left[\left.v^{\prime}\right|_{p}\right]_{p}\right)\right|_{p^{\prime}}=\left.v^{\prime}\right|_{p^{\prime}}$.

We can state a similar result if we finally project to a position that is neither a prefix nor a suffix of the original position. That is, we consider $\left.\left(v\left[v^{\prime}\right]_{p}\right)\right|_{p^{\prime}}$ and we have neither $p^{\prime} \geq p$ nor $p \geq p^{\prime}$. Note that we replaced the innermost $\left.v^{\prime}\right|_{p}$ of the previous property by $v^{\prime}$ here as this third property holds in this more general con-


Figure 4.2.4: Graphics illustrating $\left.\left(v\left[v^{\prime}\right]_{p}\right)\right|_{p^{\prime}}=\left.v\right|_{p^{\prime}}$ if $p^{\prime} \nsupseteq p$ and $p \nsupseteq p^{\prime}$
text. Therefore, the light gray triangle on the left-hand side of Figure 4.2.4 represents some arbitrary term $v^{\prime}$ instead of $\left.v^{\prime}\right|_{p}$ as it was the case before. As the graphic on the lefthand side of Figure 4.2.4 illustrates, because we project to a position incomparable with $p$, there exists some point where we leave the path defined by $p$. Thus, we end up in a part of the term $v$ that is disjunct to the light gray term. In this case it does not matter whether $v$ contains the light gray term or not. We can project to position $p^{\prime}$ in the term $v$ instead of projecting to $p^{\prime}$ in $v\left[\left.v^{\prime}\right|_{p}\right]_{p}$. The following lemma states the properties we have just illustrated.
Lemma 4.2.2: Projections and substitutions satisfy the following properties.

- For all positions $p$ and all $v \in \llbracket \tau \rrbracket$ we have $v\left[\left.v\right|_{p}\right]_{p}=v$.
- For all positions $p, p^{\prime}$ and all $v, v^{\prime} \in \llbracket \tau \rrbracket$ the following statements hold.
- If $p^{\prime} \geq p$, then we have $\left.\left(v\left[\left.v^{\prime}\right|_{p}\right]_{p}\right)\right|_{p^{\prime}}=\left.v^{\prime}\right|_{p^{\prime}}$.
- If $p^{\prime} \nsupseteq p$ and $p \nsupseteq p^{\prime}$, then we have $\left.\left(v\left[v^{\prime}\right] p\right)\right|_{p^{\prime}}=\left.v\right|_{p^{\prime}}$.

We use positions and projections to state whether a function is sequential. In the following we use $\perp$ to denote $\perp_{\tau}$ for some type $\tau$. In particular this means that we have $\perp \neq \perp$. In other words, in the following we only consider projections to valid positions, but for simplicity we do not always state this condition explicitly. The following lemma is an instance of the definition of sequentiality.
Lemma 4.2.3 (Sequential Function): In a sequential language the following holds. Let $f \in \llbracket \sigma \rightarrow \tau \rrbracket$, $p v$ be a partial value of type $\sigma$ and $r p \in \operatorname{Pos}(f p v)$ such that $\left.(f p v)\right|_{r p}=\perp$. Then, there exists $p \in$ Pos $p v$ such that $\left.p v\right|_{p}=\perp$, and for all values $p v^{\prime}$ of type $\sigma$ the following holds.

$$
\left.p v^{\prime} \sqsupseteq p v \wedge p v^{\prime}\right|_{p}=\left.\perp \Longrightarrow\left(f p v^{\prime}\right)\right|_{r p}=\perp
$$

Note that the previous lemma considers partial values $p v$ of type $\tau$, that is, elements of $\llbracket \tau \rrbracket \backslash \llbracket \tau \rrbracket \uparrow$. Therefore, a position $p$ with $\left.p v\right|_{p}=\perp$ always exists.
Definition 4.2.6 (Sequential/Non-Sequential Position): We call position $p$ of Lemma 4.2.3 a sequential position in $p v$ at result position $r p$ with respect to $f$.

We call a position $p$ a non-sequential position in $p v$ at result position $r p$ with respect to $f$ if $\left.(f p v)\right|_{r p}=\perp$ and $p \in P o s p v$ such that $\left.p v\right|_{p}=\perp$, and there exists a value $p v^{\prime}$ of type $\tau$ with the following property.

$$
\left.p v^{\prime} \sqsupseteq p v \wedge p v^{\prime}\right|_{p}=\left.\perp \wedge\left(f p v^{\prime}\right)\right|_{r p} \neq \perp
$$

Note that a position might be neither sequential nor non-sequential. However, if $p v$ is a partial value and $\left.(f p v)\right|_{r p}=\perp$, then all positions $p \in P o s p v$ with $\left.p v\right|_{p}=\perp$ are either sequential or non-sequential.

Example 4.2.5: We consider the Boolean conjunction andL from Example 2.2.1 again. We have $\left.(\llbracket a n d L \rrbracket\langle\perp, \perp\rangle)\right|_{\epsilon}=\perp$ and can apply Lemma 4.2.3. It states that there is a sequential position $p$ in $\langle\perp, \perp\rangle$ at result position $\epsilon$. Because we need $\left.\langle\perp, \perp\rangle\right|_{p}=\perp$, the only sequential position candidates are position 1 and position 2 , that is, the first and the second argument of andL. As Lemma 4.2.3 guarantees the existence of a sequential position, we have

$$
\forall v \in \llbracket \text { Bool } \rrbracket .\left.(\llbracket a n d L \rrbracket\langle v, \perp\rangle)\right|_{\epsilon}=\perp \quad \text { or } \quad \forall v \in \llbracket \text { Bool } \rrbracket .\left.(\llbracket a n d L \rrbracket\langle\perp, v\rangle)\right|_{\epsilon}=\perp .
$$

As we have $\left.(\llbracket$ and $L \rrbracket\langle$ False,$\perp\rangle)\right|_{\epsilon}=$ False $\neq \perp$, the first formula of the disjunction is false and, hence, the second formula of the disjunction is true. Thus, position 1 is a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ and positions 2 is a non-sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$.

The first argument of $a n d L$ is a sequential position in $\langle\perp, \perp\rangle$ because the function performs pattern matching on its first argument. In fact, there is a close relation between the argument positions a function performs pattern matching on and sequential positions. In the following we consider the position of an argument that is demanded by a function if we demand a specific result position. A value is demanded if it is evaluated to head normal form.

Example 4.2.6: Haskell provides a function error :: String $\rightarrow \alpha$ that takes a string and yields a run-time error if it is evaluated. If an application of error is evaluated, the execution is stopped and the corresponding string is printed. We can use this mechanism to observe, which argument of a function is demanded if we demand the result. For example, consider the evaluation of the expression error "1" \&\& error "2" in GHCi , the interpreter of the Glasgow Haskell Compiler (GHC).

```
> error "1" && error "2"
*** Exception: 1
```

As the error message " 1 " is printed, we know that \&\& demands its first argument if we demand the expression error "1" \&\& error "2". By employing exception handling we can even employ this information in Haskell itself.

If we replace the argument that is demanded by a non-erroneous value, the function might demand another argument. For example, the function \&\& demands its second argument if we demand the expression True \&\& error " 2 ".

```
> True && error "2"
*** Exception: 2
```

On the other hand, if the demanded argument is an error the function still evaluates this error independent of the other arguments. For example, if we demand the expression error "1" \&\& True the function \&\& still demands its first argument.


Figure 4.2.5: The Structure of $\llbracket$ List $\tau \rrbracket^{E r r}$

```
> error "1" && True
*** Exception: 1
```

Note that this behavior is closely connected to the existence of a sequential position. With respect to demanded positions, errors take over the role of least elements in the context of sequential positions.

In the previous examples the demand is caused by pattern matching on the corresponding argument position. A function can also demand an argument position by projecting to it. For example, consider the Haskell function const :: $\alpha \rightarrow \beta \rightarrow \alpha$, which takes two arguments and projects to its first argument. If we evaluate the expression const (error "1") (error "2") in GHCi, we can observe that const demands its first argument if we demand its result.

```
> const (error "1") (error "2")
*** Exception: 1
```

In this case the first argument is demanded because we demand the result and const projects to its first argument.

In the following, we refer to an argument position at which a value is demanded if we demand a value at a certain result position as demanded position. To give a more precise definition of demanded position and to show a connection between demanded positions and sequential positions, we sketch how the semantics, presented in Figure 2.2.4, can be extended with exceptions in the following. We denote the type and term semantics with exceptions by $\llbracket \cdot \rrbracket^{\text {Err }}$.

For simplicity we do not use strings for error messages but natural numbers. Furthermore, we do neither add a primitive to raise an exception like the Haskell function error :: String $\rightarrow \alpha$ nor a primitive to catch such exceptions as we do not need these in the following. However, adding these functions as primitives is straightforward.

We add an additional constructor Error ${ }_{\tau}$ that takes a natural number as argument to the interpretation of the Boolean type and the list types. In most cases we omit the subscript type of Error $_{\tau}$ as it is determined by the context. Figure 4.2.5 illustrates

$$
\begin{aligned}
& \llbracket \text { case } e \text { of }\left\{\text { False } \rightarrow e_{1} ; \text { True } \rightarrow e_{2}\right\} \rrbracket_{a}^{E r r}= \begin{cases}\perp & \text { if } \llbracket e \rrbracket_{a}^{E r r}=\perp \\
\text { Errorn } & \text { if } \llbracket e \rrbracket_{a}^{E r r}=\text { Error } n \\
\llbracket e_{1} \rrbracket_{a}^{E r r} & \text { if } \llbracket e \rrbracket_{a}^{E r r}=\text { False } \\
\llbracket e_{2} \rrbracket_{a}^{E r r} & \text { if } \llbracket e \rrbracket_{a}^{E r r}=\text { True }\end{cases} \\
& \llbracket \text { case } e \text { of }\left\{N i l_{\tau} \rightarrow e_{1} ; \operatorname{Cons}\left\langle x_{1}, x_{2}\right\rangle \rightarrow e_{2}\right\} \rrbracket_{a}^{E r r}
\end{aligned} \quad\left\{\begin{array}{ll}
\perp & \text { if } \llbracket e \rrbracket_{a}^{E r r}=\perp \\
\text { Errorn } n & \text { if } \llbracket e \rrbracket_{a}^{E r r}=\text { Error } n \\
\llbracket e_{1} \rrbracket_{a}^{E r r} & \text { if } \llbracket e \rrbracket_{a}^{E r r}=\text { Nil } \\
\llbracket e_{2} \rrbracket_{a\left\lfloor x_{1} \mapsto v_{1}, x_{2} \mapsto v_{2}\right]}^{E r r} & \text { if } \llbracket e \rrbracket_{a}^{E r r}=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle
\end{array}\right]
$$

Figure 4.2.6: Term Semantics for Case Expressions in a Language with Exceptions
the structure of the resulting interpretation of a list type. All exceptions are more defined than the least element $\perp_{\text {List } \tau}$, but exceptions are incomparable to each other and incomparable to the other elements of the domain. Note that $v_{1}$ as well as $v_{2}$ might be exceptions if we consider a value of the form Cons $\left\langle v_{1}, v_{2}\right\rangle$.

Basically, the exception semantics handles exceptions in the same way as it handles the least element. That is, if we perform pattern matching on an exception we yield a corresponding exception as result. However, in contrast to $\perp$, we propagate the number that is stored in an exception. Figure 4.2 .6 presents the semantics for case expressions. We have to add rules to the interpretation of case expressions that apply if the scrutinee is an exception. In this case we wrap the number that indicates the kind of exception in an Error constructor of appropriate type. For all other expressions we take the definition of $\llbracket \cdot \rrbracket$ over but replace all occurrences of $\llbracket \llbracket$ by $\llbracket \cdot \rrbracket^{E r r}$.

The following example illustrates the semantics of the functions andL and a constant function with respect to the exception semantics. Furthermore, we exemplarily show that the exception semantics models the behavior as we have observed for Haskell functions in Example 4.2.6.

Example 4.2.7: Consider the function andL from Example 2.2.1. In the exception semantics, presented in Figure 4.2.6, we get the following semantics for andL.

$$
\llbracket a n d L \rrbracket_{a}^{E r r}=\lambda\left\langle v_{1}, v_{2}\right\rangle \cdot \begin{cases}\perp & \text { if } v_{1}=\perp \\ \text { Error } n & \text { if } v_{1}=\text { Error } n \\ \text { False } & \text { if } v_{1}=\text { False } \\ v_{2} & \text { if } v_{1}=\text { True }\end{cases}
$$

If we apply $\llbracket a n d L \rrbracket^{E r r}$ to $\langle$ Error 1, Error 2$\rangle$, we get Error 1, and, if we apply $\llbracket a n d L \rrbracket^{E r r}$ to〈True, Error 2〉, we get Error 2. That is, the semantics of andL shows the same behavior we have observed for the Haskell counterpart ( \&\&) in Example 4.2.6.

Let us consider the counterpart of the Haskell function const in our simple functional language. As the simple language does not provide polymorphism, we consider the following monomorphic Boolean instance of const.

$$
\begin{aligned}
& \text { const }:: \text { Bool } \times \text { Bool } \rightarrow \text { Bool } \\
& \text { const }\langle x, y\rangle=x
\end{aligned}
$$

In the exception semantics we get the following semantics for const.

$$
\llbracket \text { const } \rrbracket_{a}^{E r r}=\lambda\left\langle v_{1}, v_{2}\right\rangle . v_{1}
$$

Thus, we have $\llbracket$ const $\rrbracket^{\text {Err }}\langle$ Error 1, Error 2$\rangle=$ Error 1, that is, const demands its first argument if we demand its result.

Definition 4.2.8 uses the exception semantics, sketched in Figure 4.2.6, to define demanded positions. Analogous to Example 4.2 . 6 we apply a function to a value where the number stored in an exception uniquely identifies its position within the value. If we apply a function to a value of this kind and get an exception at a certain result position, then the position in the argument where the exception is coming from is called a demanded position with respect to the corresponding result position.

Because we want to show a correspondence between demanded and sequential positions, we establish a relation between demanded positions and the original semantics $\llbracket \cdot \rrbracket$. Therefore, we define a logical relation that relates elements of $\llbracket \tau \rrbracket^{E r r}$ with elements of $\llbracket \tau \rrbracket$, that is, with elements that do not contain any exceptions. The idea is basically that the semantics $\llbracket \cdot \rrbracket$ and $\llbracket \cdot \rrbracket^{E r r}$ behave equally except for exceptions Errorn as these are not present in the domains of $\llbracket \rrbracket \rrbracket$. Furthermore, the semantic function $\llbracket \cdot \rrbracket^{E r r}$ treats exceptions Error $n$ in the same way as $\llbracket \cdot \rrbracket$ treats the least element $\perp$.

A logical relation (Girard 1972; Mitchell and Meyer 1985; Honsell and Sannella 1999) is a family of relations indexed by types, that is, $\left(R_{\tau}\right)_{\tau \in \text { Types }}$. For each type $\tau$ the relation $R_{\tau}$ relates two elements of the interpretation of $\tau$. The family of relations is defined inductively over the type structure. Furthermore, if $\left(R_{\tau}\right)_{\tau \in \text { Types }}$ is a logical relation, then we have $(f, g) \in R_{\sigma \rightarrow \tau}$ if and only if for all $(x, y) \in R_{\sigma}$ we have $(f x, g y) \in R_{\tau}$. In other words, two functions are related by a logical relation if and only if the functions yield related results for all related arguments.

Intuitively, the following logical relation formalizes that the semantics $\llbracket \cdot \rrbracket^{E r r}$ treats exceptions in the same way as the semantics $\llbracket \cdot \rrbracket$ treats the least element $\perp$. The logical relation $\operatorname{err}_{\tau}^{m}$ where $m$ is a natural number relates an element $v_{1}$ of $\llbracket \tau \rrbracket^{E r r}$ with an element $v_{2}$ of $\llbracket \tau \rrbracket$ if the value $v_{2}$ contains $\perp$ at all positions that contain the exception Error $m$ in $v_{1}$. Furthermore, the value $v_{2}$ contains an arbitrary value of type $\tau$ at all positions that contain an exception Error $n$ with $m \neq n$ in $v_{1}$. That is, the relation $\operatorname{err}_{m}^{\tau}$ is a kind of $\sqsubseteq$ that treats exceptions like the least element $\perp$. However, all exceptions with a specific label, namely, with label $m$, have to become $\perp$.

Definition 4.2.7 (Logical Relation $\mathrm{err}_{\tau}^{m}$ ): We define a family of logical relations indexed by a natural number and a type. For $m \in \mathbb{N}$ and $\tau \in$ Types we define
$\mathrm{err}_{\tau}^{m} \subseteq \llbracket \tau \rrbracket^{E r r} \times \llbracket \tau \rrbracket$ as follows.

$$
\begin{aligned}
\text { err }_{\text {Bool }}^{m} & =\{(\perp, b) \mid b \in \llbracket \text { Bool } \rrbracket\} \\
& \cup\{(\text { Error } n, b) \mid b \in \llbracket \text { Bool } \rrbracket, m \neq n\} \\
& \cup\{(\text { Error } m, \perp)\} \\
& \cup\{(\text { False }, \text { False }),(\text { True, True })\} \\
\text { err }_{\text {List } \tau}^{m} & =\{(\perp, l) \mid l \in \llbracket \text { List } \tau \rrbracket\} \\
& \cup\{(\text { Error } n, l) \mid l \in \llbracket \text { List } \tau \rrbracket, m \neq n\} \\
& \cup\{(\text { Error } m, \perp)\} \\
& \cup\{(\text { Nil, Nil })\} \\
& \cup\left\{(\text { Cons }\langle x, x s\rangle, \text { Cons }\langle y, y s\rangle) \mid(x, y) \in \text { err }_{\tau}^{m},(x s, y s) \in \text { err }_{\text {List }}^{m}\right\}
\end{aligned}
$$

We use the least solution of the recursive equation in the definition of $e r r_{\text {List } \tau}^{m}$.
Two tuples are related by $\operatorname{err}_{\tau_{1} \times \cdots \times \tau_{n}}^{m}$ if all their components are pointwise related by relations of corresponding types, that is, we define $\operatorname{err}_{\tau_{1} \times \cdots \times \tau_{n}}^{m}$ as follows.

$$
\operatorname{err}_{\tau_{1} \times \cdots \times \tau_{n}}^{m}=\left\{\left(\left\langle x_{1}, \ldots, x_{n}\right\rangle,\left\langle y_{1}, \ldots, y_{n}\right\rangle\right) \mid \forall i \in\{1, \ldots, n\} .\left(x_{i}, y_{i}\right) \in \operatorname{err}_{\tau_{i}}^{m}\right\}
$$

Finally, two functions are related if the images of related arguments are related, that is, we define $\operatorname{err}_{\sigma \rightarrow \tau}^{m}$ as follows.

$$
\operatorname{err}_{\sigma \rightarrow \tau}^{m}=\left\{(f, g) \mid \forall x \in \llbracket \sigma \rrbracket^{E r r}, y \in \llbracket \sigma \rrbracket \cdot(x, y) \in \operatorname{err}_{\sigma}^{m} \Longrightarrow(f x, g y) \in \operatorname{err}_{\tau}^{m}\right\}
$$

Note that the definition of $\operatorname{err}_{\sigma \rightarrow \tau}^{m}$ is determined by the demands on a logical relation. Furthermore, note that $\operatorname{err}_{\sigma \rightarrow \tau}^{m}$ relates a function that handles exceptions, that is, an element of $\llbracket \sigma \rightarrow \tau \rrbracket^{E r r}$, with a function that does not handle exceptions, that is, an element of $\llbracket \sigma \rightarrow \tau \rrbracket$.

In the following we define when we call a position a demanded position by means of the exception semantics. We use the logical relation $\operatorname{err}_{\sigma}^{n}$ to define demanded positions with respect to the original semantics $\llbracket \cdot \rrbracket$. The following definition considers some global program $P$ that is well-typed with respect to the typing rules from Section 2.2.

Definition 4.2.8 (Demanded Position): Let $f:: \sigma \rightarrow \tau \in P$ and $p v \in \llbracket \sigma \rrbracket$. If there exists $e v \in \llbracket \sigma \rrbracket^{E r r}$ and $n \in \mathbb{N}$ such that $(e v, p v) \in \operatorname{err}_{\sigma}^{n}$ and

$$
\exists!p \in \text { Posev.ev }\left.\right|_{p}=\text { Error } n
$$

as well as

$$
\left.\left(\llbracket f \rrbracket^{E r r} e v\right)\right|_{r p}=\text { Error } n,
$$

we call $p$ a demanded position in $p v \in \llbracket \sigma \rrbracket$ at result position $r p$ with respect to $f$.
In other words, we consider a partial value $e v$ that contains an exception Error $n$ at a unique position $p$. If $\llbracket f \rrbracket^{E r r}$ projects the exception Error $n$ to the result position $r p$,
then we call $p$ a demanded position at result position $r p$. The following example presents demanded positions of andL.

Example 4.2.8: Consider the exception semantics of andL from Example 4.2.7 again. We have

$$
\llbracket a n d L \rrbracket^{\text {Err }}\langle\text { Error 1, Error } 2\rangle=\text { Error } 1
$$

and

$$
(\langle\text { Error 1, Error } 2\rangle,\langle\perp, \perp\rangle) \in \operatorname{err}_{\text {Bool } \times \text { Bool }}^{1}
$$

and, thus, position 1 is a demanded position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ with respect to andL. As the logical relation $\operatorname{err}_{\text {Bool } \times \text { Bool }}^{1}$ allows us to replace occurrences of exceptions Error $n$ with $n \neq 1$ by arbitrary values of the appropriate type, we have

$$
(\langle\text { Error 1, Error } 2\rangle,\langle\perp, \text { True }\rangle) \in \operatorname{err}_{\text {Bool } \times \text { Bool }}^{1} \text {. }
$$

Thus, position 1 is also a demanded position in $\langle\perp$, True $\rangle$ at result position $\epsilon$ with respect to andL. Furthermore, we have

$$
\llbracket a n d L \rrbracket^{\text {Err }}\langle\text { True, Error } 2\rangle=\text { Error } 2
$$

and

$$
\langle\text { True, Error } 2\rangle,\langle\text { True }, \perp\rangle) \in \text { err }_{\text {Bool } \times \text { Bool }}^{2}
$$

and, thus, position 2 is a demanded position in $\langle$ True, $\perp\rangle$ at result position $\epsilon$ with respect to andL.

To prove a connection between demanded and sequential positions we use two utilities. The first utility is called the Basic Lemma or Fundamental Theorem in the context of logical relations. Mostly, logical relations are defined for a simply typed lambda calculus with an explicit fixpoint combinator. In this setting the following lemma simply holds for all logical relations. As we do not consider a lambda calculus but a language with data structures and explicit recursion, we prove the following lemma in Appendix A.

We consider an expression $e$ of some type $\tau$ and the interpretation of $e$ with respect to environments $a_{1}$ and $a_{2}$, that is, $\llbracket e \rrbracket_{a_{1}}^{E r r}$ and $\llbracket e \rrbracket_{a_{2}}$. The lemma states that these interpretations are related by $\operatorname{err}_{\tau}^{m}$ if the images of $a_{1}$ and $a_{2}$ are related by $\operatorname{err}_{\tau^{\prime}}^{m}$ for all variables. Here the type $\tau^{\prime}$ is the type of the corresponding variable. We use the following lemma to prove that every demanded position is also a sequential position.
Lemma 4.2.4: Assume $\Gamma \vdash e:: \tau$ and let $a_{1}$ and $a_{2}$ be environments such that for all $x:: \tau^{\prime}$ in $\Gamma$ we have $\left(a_{1}(x), a_{2}(x)\right) \in \operatorname{err}_{\tau^{\prime}}^{m}$. In this case we have $\left(\llbracket e \rrbracket \rrbracket_{a_{1}}^{E r r}, \llbracket e \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau}^{m}$.

The previous lemma in particular states that for a function $f:: \sigma \rightarrow \tau$ we have $\left(\llbracket f \rrbracket_{a}^{E r r}, \llbracket f \rrbracket_{a}\right) \in e r r_{\sigma \rightarrow \tau}^{m}$ for all $m \in \mathbb{N}$. By means of this statement we get a connection between the interpretation of a function with respect to the two semantics. In addition, we employ a connection between $\mathrm{err}_{\tau}^{m}$ and the projection defined in Definition 4.2.4. Therefore, we extend projections from $\llbracket \tau \rrbracket$ to $\llbracket \tau \rrbracket^{E r r}$ by handling values
of the form Error $n$ in the same way as we handle the least element $\perp$. By means of this extension of the projections we can characterize values $e v \in \llbracket \tau \rrbracket^{E r r}$ and $p v \in \llbracket \tau \rrbracket$ with $(e v, p v) \in \operatorname{err}_{\tau}^{m}$ as follows. For all positions $p$ we have

$$
\left.e v\right|_{p}=\text { Error }\left.m \Longrightarrow p v\right|_{p}=\perp
$$

as well as

$$
m \neq\left. n \wedge e v\right|_{p}=\text { Error }\left.n \Longrightarrow p v\right|_{p} \sqsupseteq \perp .
$$

These implications provide a kind of formalisation of the intuitive interpretation of the logical relation $\mathrm{err} r_{\tau}^{m}$. In other words, while we may replace occurrences of the exception Error m in ev by $\perp$, we may replace occurrences of other exceptions Error $n$ with $m \neq n$ by arbitrary values.

On the basis of these observations we can show that every demanded position in $p v$ at result position $r p$ with respect to $f$ is a sequential position in $p v$ at result position $r p$ with respect to $\llbracket f \rrbracket$. Note that we have defined demanded positions with respect to a syntactic function $f:: \sigma \rightarrow \tau$ while we have defined sequential positions with respect to a semantic function $f \in \llbracket \sigma \rightarrow \tau \rrbracket$.

Lemma 4.2.5: Let $f:: \sigma \rightarrow \tau \in P$. If $p$ is a demanded position in $p v$ at result position $r p$ with respect to $f$, then $p$ is a sequential position in $p v$ at result position $r p$ with respect to $\llbracket f \rrbracket$.

The previous lemma shows that the set of demanded positions is a subset of the set of sequential positions. However, there are sequential positions that are not demanded positions. While with respect to some value at every result position there is always only a single demanded position, there might be several sequential positions at the same result position.

Example 4.2.9: Consider the following definition of a projection to the second argument that is strict in its first argument.

$$
\begin{aligned}
& \text { strictSnd }:: \text { Bool } \times \text { Bool } \rightarrow \text { Bool } \\
& \text { strictSnd }\langle x, y\rangle= \\
& \text { case } x \text { of } \\
& \text { False } \rightarrow y \\
& \text { True } \rightarrow y
\end{aligned}
$$

We get the following semantics for strictSnd with respect to the original semantics. Note that we have combined the cases $v_{1}=$ False and $v_{2}=$ True to a single case.

$$
\llbracket s t r i c t S n d \rrbracket=\lambda\left\langle v_{1}, v_{2}\right\rangle \cdot \begin{cases}\perp & \text { if } v_{1}=\perp \\ v_{2} & \text { otherwise }\end{cases}
$$

With respect to this semantics we have

$$
\left.(\llbracket \text { strictSnd } \rrbracket\langle\perp, b\rangle)\right|_{\epsilon}=\perp
$$

for all values $b$ of type Bool. Likewise, we have

$$
\left.(\llbracket s t r i c t S n d \rrbracket\langle b, \perp\rangle)\right|_{\epsilon}=\perp
$$

for all values $b$ of type Bool. Thus, position 1 as well as position 2 are sequential positions in $\langle\perp, \perp\rangle$ at position $\epsilon$ with respect to strictSnd.

Now, let us consider the exception semantics of strictSnd.

$$
\llbracket s t r i c t S n d \rrbracket^{E r r}=\lambda\left\langle v_{1}, v_{2}\right\rangle \cdot \begin{cases}\perp & \text { if } v_{1}=\perp \\ \text { Errorn } n & \text { if } v_{1}=\text { Error } n \\ v_{2} & \text { otherwise }\end{cases}
$$

With respect to this semantics we have

$$
\llbracket \text { strictSnd } \rrbracket^{\text {Err }}\langle\text { Error 1, Error } 2\rangle=\text { Error } 2
$$

and

$$
\left.(\langle\text { Error 1, Error } 2\rangle,\langle\perp, \perp\rangle) \in \text { err }_{\langle\text {Bool }, \text { Bool }\rangle}^{2}\right\rangle
$$

Therefore, position 2 is a demanded position in $\langle\perp, \perp\rangle$ at position $\epsilon$ with respect to strictSnd.

The exception semantics presented here does not match the semantics of Haskell exceptions (Peyton Jones et al. 1999). The exceptions employed by Haskell are called imprecise exceptions. The semantics of these exceptions is less restrictive to guarantee that certain transformations are semantic-preserving. The basic idea behind imprecise exceptions is to use sets of exceptions. More precisely, instead of a single natural number an exception contains a set of natural numbers where the set denotes the set of possible exceptions that might be raised by the evaluation of an expression. For example, consider the Haskell function head, which is defined as follows.

```
head :: \([\alpha] \rightarrow \alpha\)
head [] = error "empty list"
head \(\left(x:_{\_}\right)=x\)
```

If we apply head to an exception like error "argument", in the imprecise exceptions semantics we get a set containing the exception "argument" as well as the exception "empty list". If we catch the exception by employing the Haskell function catch, it choses one of the possible exceptions non-deterministically. Therefore, catch is in the IO monad as it can be considered to ask some kind of oracle, which exception to choose.

As long as the compiler does not apply any optimizations, in practice, Haskell exceptions behave in the way specified by the exception semantics $\llbracket \cdot \rrbracket^{E r r}$. In the same manner as we assume a simplified exception semantics, the behavior of the black-box testing tool Lazy Smallcheck by Runciman et al. (2008) is based on this simplified exception model. Lazy Smallcheck employs exception handling mecha-
nisms to implement a kind of needed narrowing (Antoy et al. 1994) to systematically generate test cases for property-based testing.

We will reconsider demanded positions when we present the implementation of Sloth. There, we are interested in demanded positions because they are an easy way to identify sequential positions of a function. For now, we switch to the semantics $\llbracket \cdot \rrbracket$ again and deal with the question when a function is minimally strict.

### 4.3 Minimally Strict Functions

In this section we employ the definition of sequentiality, presented in the previous section, to define a criterion to check whether a sequential function is minimally strict. When we only consider sequential functions, we are looking for minimally strict functions and not for a least strict function.

We start by providing a formal definition of "minimally strict".
Definition 4.3.1: A sequential function $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ is minimally strict if for all sequential functions $g \in \llbracket \sigma \rightarrow \tau \rrbracket$ the relation $g \preceq f$ implies $g=f$.

In order to provide a criterion that states whether a function is minimally strict we start with some preliminaries. By $\Sigma_{C}$ we denote the set of semantic constructor symbols. As we only consider Booleans and lists as algebraic data types, we have $\Sigma_{C}=\{$ False, True, Nil,Cons $\}$. The refinement of a partial value is the replacement of a subterm that is $\perp$ by a partial value of the form $C \perp$ where $C$ is a constructor from $\Sigma_{C}$. We consider two kinds of refinements: a refinement at a sequential position and a refinement at a non-sequential position.
Definition 4.3.2 (Refinement Relations): For all values $v$ of type $\tau$ and all positions $p \in \operatorname{Pos} v$ with $\left.v\right|_{p}=\perp$ and $\left.v\right|_{p} \in \llbracket \tau^{\prime} \rrbracket$ we define the refinement of $v$ as follows. We have $v \lessdot_{p} v^{\prime}$ if there exists $C \in \Sigma_{C}$ such that $C \in \llbracket \sigma \rightarrow \tau^{\prime} \rrbracket$ and $v^{\prime}=v[C \perp]_{p}$. Note that, if $C$ takes multiple arguments, then $\sigma=\tau_{1} \times \cdots \times \tau_{n}$ and $\perp=\left\langle\perp_{\tau_{1}}, \ldots, \perp_{\tau_{n}}\right\rangle$.

On basis of this definition and the notion of sequential and non-sequential positions we define two refinement relations. We relate values $v$ and $v^{\prime}$ of type $\tau$ and a result position $r p$ for a given function $f$ by

$$
v \mapsto_{r p}^{f} v^{\prime}
$$

if there exists $p \in \operatorname{Pos} v$ such that $v \lessdot_{p} v^{\prime}$ and $p$ is a sequential position in $v$ at result position $r p$ with respect to $f$. We also refer to this as a sequential step. Furthermore we relate values $v$ and $v^{\prime}$ of type $\tau$ and a result position $r p$ for a given function $f$ by

$$
v \rightsquigarrow f{ }_{r p}^{f} v^{\prime}
$$

if there exists $p \in \operatorname{Pos} v$ such that $v \lessdot_{p} v^{\prime}$ and $p$ is a non-sequential position in $v$ at result position $r p$ with respect to $f$. We also refer to this as a non-sequential step.

Whenever $f$ is determined by the context, we omit it. Furthermore, although $f$ is supposed to be a semantic function for simplicity we annotate the refinement relations with the corresponding syntactic function.

In the following, if we have $v \lessdot_{p} v^{\prime}$ but disregard the refinement position, we omit the position and write $v \lessdot v^{\prime}$. In the more general setting of an arbitrary partially ordered set $(P, \sqsubseteq)$ for elements $x \in P$ and $y \in P$ the relation $\lessdot$ is called covering relation and $x \lessdot y$ is defined as $x \sqsubset y$ and there exists no $z \in P$ with $x \sqsubset z \sqsubset y$.

Let us consider some values that are related by one of the refinement relations.

Example 4.3.1: Consider andL from Example 2.2.1 again. We have

$$
\langle\perp, \perp\rangle \mapsto_{\epsilon}^{\text {andL }}\langle\text { False }, \perp\rangle
$$

as well as

$$
\langle\perp, \perp\rangle \longmapsto_{\epsilon}^{\text {andL }}\langle\text { True, } \perp\rangle
$$

because position 1 is a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ with respect to andL. Furthermore, we have

$$
\langle\text { True, } \perp\rangle \mapsto_{\epsilon}^{a n d L}\langle\text { True, False }\rangle
$$

as well as

$$
\langle\text { True, } \perp\rangle \longmapsto_{\epsilon}^{\text {andL }}\langle\text { True, True }\rangle
$$

because position 2 is a sequential position in $\langle$ True, $\perp\rangle$ at position $\epsilon$ with respect to andL. On the other hand we have

$$
\langle\perp, \perp\rangle \rightsquigarrow_{\epsilon}^{\text {andL }}\langle\perp, \text { False }\rangle
$$

as well as

$$
\langle\perp, \perp\rangle \rightsquigarrow_{\epsilon}^{a n d L}\langle\perp, \text { True }\rangle
$$

because position 2 is a non-sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ with respect to andL. Note that we neither have

$$
\langle\text { False, } \perp\rangle \mapsto_{\epsilon}^{\text {andL }}\langle\text { False, False }\rangle
$$

nor

$$
\langle\text { False, } \perp\rangle \rightsquigarrow_{\epsilon}^{a n d L}\langle\text { False, False }\rangle
$$

because $\left.(\llbracket$ and $L \rrbracket\langle$ False, $\perp\rangle)\right|_{\epsilon}=$ False $\neq \perp$, and, therefore, position 2 is neither a sequential nor a non-sequential position in $\langle$ False, $\perp\rangle$ at result position $\epsilon$ with respect to andL.

If we consider the function and from Example 4.1.2, we have

$$
t_{1}{ }_{\epsilon}^{\text {and }} t_{2}
$$

for all $t_{1}, t_{2} \in \llbracket$ Bool $\times$ Bool $\rrbracket$ with $t_{1} \lessdot t_{2}$ because $\llbracket a n d \rrbracket$ yields $\perp$ as long as any of its arguments is still undefined. In other words, all refinement steps with respect to $\llbracket a n d \rrbracket$ are sequential steps.

On basis of the definition of refinement steps we define the characteristic set of a function. The elements of the characteristic set are test cases used to check whether a function is minimally strict. The characteristic set of a function $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ is a subset of the set of all pairs of argument values and valid result positions such that $f$ applied to the argument yields $\perp$ at the result position. That is, the characteristic set
is a subset of the set $\left\{(v, r p)|v \in \llbracket \sigma \rrbracket, r p \in \operatorname{Pos}(f v),(f v)|_{r p}=\perp\right\}$. We omit pairs from the characteristic set for which the behavior of $f$ is determined by sequentiality.
Definition 4.3.3 (Characteristic Set): The characteristic set $C_{f}$ of a sequential function $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ is inductively defined as follows.

1. For all result positions $r p$ such that $\left.(f \perp)\right|_{r p}=\perp$ we have $(\perp, r p) \in C_{f}$.
2. For all pairs $(v, r p) \in C_{f}$ and all values $v^{\prime}$ such that $v \mapsto_{r p}^{f} v^{\prime}$ and all result positions $r p^{\prime}$ such that $\left.\left(f v^{\prime}\right)\right|_{r p^{\prime}}=\perp$ we have $\left(v^{\prime}, r p^{\prime}\right) \in C_{f}$.

Obviously, we have $C_{f} \subseteq\left\{(v, r p)|v \in \llbracket \sigma \rrbracket, r p \in \operatorname{Pos}(f v),(f v)|_{r p}=\perp\right\}$. Furthermore, if we have $(v, r p) \in C_{f}$, then there exists a sequence of sequential steps $\perp \hookrightarrow^{*} v$ that is monotonic with respect to the considered result positions. Thus, if we have

$$
\perp \mapsto^{*} v_{1} \mapsto_{r p} v_{2} \mapsto^{*} v_{3} \mapsto_{r p^{\prime}} v_{4} \mapsto^{*} v
$$

then $r p$ is a prefix of $r p^{\prime}$, denoted by $r p \leq r p^{\prime}$.
Example 4.3.2: Let us consider the Boolean conjunctions andL from Example 2.2.1 as well as and from Example 4.1.2, again. What do the characteristic sets for $\llbracket a n d L \rrbracket$ and【and】 look like?
In the following, for simplicity, we say that "a value $v$ is not in $C_{f}$ " or " $v$ is not an element of $C_{f}{ }^{\prime \prime}$ if we have $(v, r p) \notin C_{f}$ for all result positions $r p$.

First, we consider the function $\llbracket a n d L \rrbracket$. We have $\left.(\llbracket a n d L \rrbracket\langle\perp, \perp\rangle)\right|_{\epsilon}=\perp$, and, by definition of the characteristic set, we have $(\langle\perp, \perp\rangle, \epsilon) \in C_{\llbracket a n d L \rrbracket}$. Note that $\langle\perp, \perp\rangle$ is the least element of $\llbracket \mathrm{Bool} \times \mathrm{Bool} \rrbracket$, which is denoted by $\perp$ in Definition 4.3.3. Position $\epsilon$ is the only valid result position of $\llbracket a n d L \rrbracket\langle\perp, \perp\rangle$, which yields $\perp$, when evaluated. Thus, the value $\langle\perp, \perp\rangle$ is not in $C_{\llbracket a n d L \rrbracket}$.

As we have observed in Example 4.3.1, we have

$$
\langle\perp, \perp\rangle \mapsto_{\epsilon}^{\text {andL }}\langle\text { False }, \perp\rangle
$$

as well as

$$
\langle\perp, \perp\rangle \mapsto_{\epsilon}^{a n d L}\langle\text { True }, \perp\rangle
$$

Nevertheless, as $\llbracket a n d L \rrbracket\langle$ False, $\perp\rangle$ yields a total result, the value $\langle$ False, $\perp\rangle$ is not in $C_{\llbracket a n d L \rrbracket}$ because there exists no result position $r p$ such that $\left.(\llbracket a n d L \rrbracket\langle$ False, $\perp\rangle)\right|_{r p}=$ $\perp$. That is, some values $v$ are not in the characteristic set because the considered function yields a total result if it is applied to $v$. In contrast, the pair $(\langle$ True, $\perp\rangle, \epsilon)$ is in the characteristic set as we have $\left.(\llbracket$ andL $\rrbracket\langle$ True, $\perp\rangle)\right|_{\epsilon}=\perp$.

In Example 4.3.1 we have observed that

$$
\langle\perp, \perp\rangle \rightsquigarrow_{\epsilon}^{a n d L}\langle\perp, \text { False }\rangle
$$

as well as

$$
\langle\perp, \perp\rangle \rightsquigarrow_{\epsilon}^{a n d L}\langle\perp, \text { True }\rangle
$$

holds and, thus, $\langle\perp$, False $\rangle$ and $\langle\perp$, True $\rangle$ are not in the characteristic set. These values are not in the characteristic set of $\llbracket a n d L \rrbracket$ because the behavior for these arguments
is determined by sequentiality．More precisely，as position 1 is a sequential position in $\langle\perp, \perp\rangle$ we have $\llbracket a n d L \rrbracket\langle\perp$, False $\rangle=\perp$ and $\llbracket a n d L \rrbracket\langle\perp$, True $\rangle=\perp$ by sequentiality．

The value $\langle$ False，False $\rangle$ is not in the characteristic set because neither $\langle$ False，$\perp\rangle$ nor $\langle\perp$, False $\rangle$ are in the characteristic set．We do not have to consider $\langle$ False，False $\rangle$ because it is more defined than $\langle$ False，$\perp\rangle$ and andL already yields a total result for $\langle$ False，$\perp\rangle$ ．Therefore，all values more defined than $\langle$ False，$\perp\rangle$ are not in the charac－ teristic set．

As we have observed in Example 4．3．1，we have the sequential steps

$$
\langle\text { True, } \perp\rangle \mapsto_{\epsilon}^{\text {andL }}\langle\text { True, False }\rangle
$$

and

$$
\langle\text { True, } \perp\rangle \mapsto_{\epsilon}^{\text {andL }}\langle\text { True, True }\rangle .
$$

However，as the results of $\llbracket a n d L \rrbracket$ for $\langle$ True，False〉 as well as for $\langle$ True，True〉 are total we have neither $\langle$ True，False $\rangle \in C_{\llbracket a n d L \rrbracket}$ nor $\langle$ True，True $\rangle \in C_{\llbracket a n d L \rrbracket}$ ．

In summary，we get the following characteristic set for 【andL】】．

$$
C_{\llbracket a n d L \rrbracket}=\{(\langle\perp, \perp\rangle, \epsilon),(\langle\text { True }, \perp\rangle, \epsilon)\}
$$

The value $\langle$ False，$\perp\rangle$ is missing in $C_{\llbracket a n d L \rrbracket}$ because andL yields a total value，namely， False，for this argument．The arguments $\langle\perp$, False $\rangle$ and $\langle\perp$, True $\rangle$ are missing in $C_{\llbracket a n d L \rrbracket}$ because there is no sequence of sequential steps from $(\langle\perp, \perp\rangle, \epsilon)$ to these values．

In the case of $\llbracket a n d \rrbracket$ all steps are sequential with respect to the result position $\epsilon$ ．

$$
C_{\llbracket \text { and } \rrbracket}=\left\{\begin{array}{c}
(\langle\perp, \perp\rangle, \epsilon),(\langle\text { False }, \perp\rangle, \epsilon),(\langle\text { True }, \perp\rangle, \epsilon), \\
(\langle\perp, \text { False }\rangle, \epsilon),(\langle\perp, \text { True }\rangle, \epsilon) .
\end{array}\right\} .
$$

Therefore，the characteristic set consists of all pairs $(v, \epsilon)$ where $v$ is a value of type Bool $\times$ Bool for which $\llbracket a n d \rrbracket$ yields a non－total result．

By means of the characteristic set we define a criterion to check whether a function is minimally strict．The criterion states that a sequential function $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ is not minimally strict if and only if there exists $(v, r p) \in C_{f}$ with $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$ ．In the remainder of this chapter we prove the following theorem，where $\inf _{f} v$ is defined by $\inf _{f} v=\Pi\{f t v \mid t v \in \llbracket \sigma \rrbracket \uparrow, v \sqsubseteq t v\}$（see Definition 4．1．3）．

Theorem 4．3．1：A sequential function $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ is minimally strict if and only if we have $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$ for all $(v, r p) \in C_{f}$ ．

In the following two sections（Section 4．3．1 and Section 4．3．2）we assume that all considered functions are sequential．

### 4.3.1 Sufficiency of the Criterion

In this section we show that $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$ for all $(v, r p) \in C_{f}$ is a sufficient condition for $f$ being minimally strict. In other words, we show that $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$ for all $(v, r p) \in C_{f}$ implies that $f$ is minimally strict.
To prove this statement we show that the characteristic set of a function is in a certain sense characteristic. We consider two functions $f$ and $g$ such that $f \preceq g$. If we have $\left.(f v)\right|_{r p}=\perp$ for all values $v \in \llbracket \sigma \rrbracket$ and valid result positions $r p$ such that $\left.(g v)\right|_{r p}=\perp$, then we have $f=g$. In the following we show that we do not have to check all values $v$ and results positions $r p$ such that $\left.(g v)\right|_{r p}=\perp$. We already have $f=g$ if $\left.(f v)\right|_{r p}=\perp$ for all $(v, r p) \in C_{g}$.

The characteristic set contains only finite values. Therefore, $\left.(f v)\right|_{r p}=\perp$ for all $(v, r p) \in C_{g}$ merely implies that $f$ and $g$ agree for all finite values $v$. Nevertheless, as we use algebraic cpos the behavior of continuous functions for non-finite values is determined by their behavior for finite values. Consider two functions $f, g \in$ $\llbracket \sigma \rightarrow \tau \rrbracket$ and a non-finite value $v \in \llbracket \sigma \rrbracket$. If $f$ and $g$ agree for all finite values, they also agree for all non-finite values as the following equational reasoning shows.

$$
\begin{aligned}
f v & =f \bigsqcup\left\{v^{\prime} \mid v^{\prime} \in \llbracket \sigma \rrbracket, v^{\prime} \sqsubseteq v, v^{\prime} \text { finite }\right\} & & \text { algebraicity } \\
& =\bigsqcup\left\{f v^{\prime} \mid v^{\prime} \in \llbracket \sigma \rrbracket, v^{\prime} \sqsubseteq v, v^{\prime} \text { finite }\right\} & & \text { continuity of } f \\
& =\bigsqcup\left\{g v^{\prime} \mid v^{\prime} \in \llbracket \sigma \rrbracket, v^{\prime} \sqsubseteq v, v^{\prime} \text { finite }\right\} & & f \text { and } g \text { agree for finite values } \\
& =g \bigsqcup\left\{v^{\prime} \mid v^{\prime} \in \llbracket \sigma \rrbracket, v^{\prime} \sqsubseteq v, v^{\prime} \text { finite }\right\} & & \text { continuity of } g \\
& =g v & & \text { algebraicity }
\end{aligned}
$$

As a first step towards the sufficiency of the criterion, we show that for all pairs $(v, r p)$ with $\left.(f v)\right|_{r p}=\perp$ that are not in the characteristic set there exists a pair $\left(v^{\prime}, r p\right)$ that is in the characteristic set. Furthermore, there exists a sequence of non-sequential steps at position $r p$ from $v^{\prime}$ to $v$, that is, $v^{\prime}\left(\rightsquigarrow_{r p}\right)^{*} v$. Figure 4.3.1 illustrates the statement. The white rectangle represents the set of all pairs $(v, r p)$ such that $\left.(f v)\right|_{r p}=\perp$ while the gray rectangle represents the characteristic set $C_{f}$. For every pair $(v, r p)$ that is not in the characteristic set there exists a pair $\left(v^{\prime}, r p\right)$ in the characteristic set and a non-sequential path from $v^{\prime}$ to $v$. Considering this sequence, we then employ the following lemma to show that we have $\left.(f v)\right|_{r p}=\perp$. In the following we assume that $\cdot[v]_{p}$ has a higher precedence than function application, that is, $f v\left[v^{\prime}\right]_{p}$ stands for $f\left(v\left[v^{\prime}\right]_{p}\right)$.
Lemma 4.3.1: Let $p$ be a non-sequential position of $f$ in $v$ at result position $r p$. Then for all values $v^{\prime}$ of appropriate type we have $\left.\left(f v\left[v^{\prime}\right]_{p}\right)\right|_{r p}=\perp$.
Note that the previous lemma implies that we have $\left.\left(f v_{2}\right)\right|_{r p}=\perp$ if $v_{1} \rightsquigarrow_{r p} v_{2}$ because we have $v_{2}=v_{1}[C \perp]_{p}$ for some constructor $C$ and a non-sequential position $p$. Thus, if there exists a non-sequential sequence $v^{\prime}\left(\rightsquigarrow_{r p}\right)^{*} v$ we have $\left.(f v)\right|_{r p}=\perp$.

To prove that there always exists a non-sequential sequence from a pair that is in the characteristic set to one that is not, we consider an arbitrary pair $(v, r p) \notin C_{f}$ with $\left.(f v)\right|_{r p}=\perp$. The following lemma shows that $v \neq \perp$ implies that there exists a predecessor $v^{\prime}$ in the sense that we have $v^{\prime} \rightsquigarrow_{r p} v$ or we have $v^{\prime} \mapsto_{r p^{\prime}} v$ and $r p^{\prime} \leq r p$.


Figure 4.3.1: Illustration of Lemma 4.3.5

Lemma 4.3.2: For all values $v \in \llbracket \tau \rrbracket \backslash\{\perp\}$ and $r p \in \operatorname{Pos}(f v)$ with $\left.(f v)\right|_{r p}=\perp$ there exists $v^{\prime} \in \llbracket \tau \rrbracket$ and $r p^{\prime} \in \operatorname{Pos}\left(f v^{\prime}\right)$ such that $v^{\prime} \rightsquigarrow_{r p} v$ or such that $v^{\prime} \rightarrow_{r p^{\prime}} v$ and $r p^{\prime} \leq r p$.

Let us get back to the proof of the statement illustrated by Figure 4.3.1. We only consider finite values $v$ and prove the statement by induction over the number of constructors in $v$. As mentioned before, we consider a pair $(v, r p) \notin C_{f}$ with $\left.(f v)\right|_{r p}=\perp$. By the previous lemma there exists a value $v^{\prime}$ such that we have either $v^{\prime} \rightsquigarrow_{r p} v$ or we have $v^{\prime} \mapsto_{r p^{\prime}} v$ and $r p^{\prime} \leq r p$. First we look at the former case. If we have $\left(v^{\prime}, r p\right) \in C_{f}$, we are finished as we have a non-sequential sequence from $v^{\prime}$ to $v$. Therefore, we consider the case that $\left(v^{\prime}, r p\right) \notin C_{f}$. By using the induction hypothesis we get a non-sequential sequence that arrives at $v^{\prime}$. In other words, there exists $\left(v^{\prime \prime}, r p\right) \in C_{f}$ such that $v^{\prime \prime}\left(\rightsquigarrow_{r p}\right)^{*} v^{\prime}$. Thus, we use $v^{\prime} \rightsquigarrow_{r p} v$ to construct a new non-sequential sequence that arrives at $v$.

Let us look at the remaining case, namely, $v^{\prime} \rightarrow_{r p^{\prime}} v$ with $r p^{\prime} \leq r p$. As we have $(v, r p) \notin C_{f}$, we get $\left(v^{\prime}, r p^{\prime}\right) \notin C_{f}$ by definition of the characteristic set. By induction hypothesis we conclude that there exists $\left(v^{\prime \prime}, r p^{\prime}\right) \in C_{f}$ such that $v^{\prime \prime}\left(\rightsquigarrow_{r p^{\prime}}\right)^{*} v^{\prime} \mapsto_{r p^{\prime}} v$. As we are looking for a purely non-sequential sequence, in the following, we show that we can rearrange this sequence such that all sequential steps are "performed" before any non-sequential steps. Before we prove the statement for arbitrary sequences, we consider sequences of length two, that is, we have a sequence of the form $v_{1} \rightsquigarrow r p v_{2} \mapsto_{r p} v_{3}$.

Example 4.3.3: Consider andL from Example 2.2.1 again and the following sequence of a non-sequential and a sequential step.

$$
\langle\perp, \perp\rangle \rightsquigarrow_{\epsilon}^{\text {andL }}\langle\perp, \text { True }\rangle \longmapsto_{\epsilon}^{\text {andL }}\langle\text { True, True }\rangle .
$$

If we interchange the steps, we get the following sequence of two sequential steps.

$$
\langle\perp, \perp\rangle \mapsto_{\epsilon}^{a n d L}\langle\text { True, } \perp\rangle \mapsto_{\epsilon}^{a n d L}\langle\text { True, True }\rangle
$$

Now consider the following sequence.

$$
\langle\perp, \perp\rangle \rightsquigarrow_{\epsilon}^{\text {andL }}\langle\perp, \text { False }\rangle \mapsto_{\epsilon}^{\text {andL }}\langle\text { False, False }\rangle .
$$

If we first perform the sequential step, the step remains sequential.

$$
\langle\perp, \perp\rangle \mapsto_{\epsilon}^{\text {andL }}\langle\text { False }, \perp\rangle
$$

However, we cannot perform a second step as we have $\left.(\llbracket$ and $\rrbracket \rrbracket\langle$ False, $\perp\rangle)\right|_{\epsilon}=$ False. Thus, there exists neither a sequential nor a non-sequential position in $\langle$ False, $\perp\rangle$. In other words, we have neither

$$
\langle\text { False, } \perp\rangle \mapsto_{\epsilon}^{a n d L}\langle\text { False, False }\rangle
$$

nor

$$
\langle\text { False }, \perp\rangle \rightsquigarrow_{\epsilon}^{a n d L}\langle\text { False, False }\rangle .
$$

We have observed that a non-sequential step might become sequential or might vanish if we interchange it with a sequential step. Finally, we show that the non-sequential step might even remain non-sequential. We define the following generalization of andL to three Boolean arguments, called andL3.

```
andL3 \(::\) Bool \(\times\) Bool \(\times\) Bool \(\rightarrow\) Bool
\(\operatorname{andL} 3\langle x, y, z\rangle=\)
    case \(x\) of
        False \(\rightarrow\) False
        True \(\rightarrow\) case \(y\) of
        False \(\rightarrow\) False
        True \(\rightarrow z\)
```

Employing the semantics of Figure 2.2.4 we get the following semantics for andL3.

$$
\llbracket \text { andL3』 } \rrbracket \lambda\left\langle v_{1}, v_{2}, v_{3}\right\rangle \cdot \begin{cases}\perp & \text { if } v_{1}=\perp \\ \text { False } & \text { if } v_{1}=\text { False } \\ \perp & \text { if } v_{1}=\text { True } \wedge v_{2}=\perp \\ \text { False } & \text { if } v_{1}=\text { True } \wedge v_{2}=\text { False } \\ v_{3} & \text { otherwise }\end{cases}
$$

Now consider the following sequence of a non-sequential and then a sequential step.

$$
\langle\perp, \perp, \perp\rangle \rightsquigarrow_{\epsilon}^{a n d L 3}\langle\perp, \perp, \text { True }\rangle \mapsto_{\epsilon}^{a n d L 3}\langle\text { True }, \perp, \text { True }\rangle
$$

If we interchange these steps, we get a sequential and then a non-sequential step.

$$
\langle\perp, \perp, \perp\rangle \mapsto_{\epsilon}^{\text {andL3 } 3}\langle\text { True }, \perp, \perp\rangle \rightsquigarrow_{\epsilon}^{\text {andL3 }}\langle\text { True }, \perp, \text { True }\rangle
$$

To show that the second step is non-sequential we have to provide a value $v$ with $v \sqsupseteq\langle$ True, $\perp, \perp\rangle$ and $\left.v\right|_{3}=\perp$ such that $\left.(\llbracket a n d L 3 \rrbracket v)\right|_{\epsilon} \neq \perp$. The value $\langle$ True, False, $\perp\rangle$ satisfies these conditions because we have $\left.(\llbracket a n d L 3 \rrbracket\langle$ True, False, $\perp\rangle)\right|_{\epsilon}=$ False .

As a next step, we prove that, if there exists a sequence of a non-sequential and a final sequential step, then there exists a sequence of a sequential and an arbitrary step that arrives at the same value as the original sequence. Let us consider a sequence of a non-sequential and a sequential step, namely, a sequence of the following form.

$$
v_{1} \rightsquigarrow r p_{1} v_{2} \mapsto_{r p_{2}} v_{3}
$$

In fact, we do not consider arbitrary sequences of this form, but only sequences that are monotonic with respect to result positions, that is, we have $r p_{1} \leq r p_{2}$. In this case by Lemma 4.3.1 we have $\left.\left(f v_{2}\right)\right|_{r p_{1}}=\perp$ and, therefore, $r p_{1}=r p_{2}$. Thus, we can even only consider a particular form of these sequences, namely, the following.

$$
v_{1} \rightsquigarrow r p_{1} v_{2} \mapsto_{r p_{1}} v_{3}
$$

As Example 4.3 .3 shows, we do not always arrive at the result value of the original sequence if we consider sequences of this form. That is, if we first perform the sequential step we might not be able to arrive at the final value by neither a sequential nor a non-sequential step. Therefore, we only consider sequences such that $\left.\left(f v_{3}\right)\right|_{r p_{3}}=\perp$ for some result position $r p_{3}$ with $r p_{1} \leq r p_{3}$. In this case we have $\left.(f v)\right|_{r p_{2}}=\perp$ for all values $v$ with $v_{1} \sqsubseteq v \sqsubseteq v_{3}$ and some result position $r p_{2}$ with $r p_{1} \leq r p_{2} \leq r p_{3}$. This follows from the definition of the relation $\sqsubseteq$ of the considered domains as well as monotonicity of $f$. As we have $\left.(f v)\right|_{r p_{2}}=\perp$, there either exists a sequential or a non-sequential step that arrives at $v_{3}$.

Now that we know that we always arrive at the final value, let us consider a sequence of the following form.

$$
v_{1} \rightsquigarrow{ }_{r p_{1}} v_{2} \mapsto_{r p_{1}} v_{3}
$$

If we first perform the sequential step of this sequence, we have to distinguish two cases. In the first case we get a sequential step from $v_{1}$ to some value $v$ and a non-sequential step from $v$ to $v_{2}$. As the latter step is non-sequential and by precondition we have $\left.\left(f v_{2}\right)\right|_{r p_{3}}=\perp$, we get $\left.(f v)\right|_{r p_{3}}=\perp$ because non-sequential steps do not produce more defined function results (Lemma 4.3.1). In the other case the resulting steps are both sequential. In this case the second step might be at a result position $r p_{2}$ that is between $r p_{1}$ and $r p_{3}$. In other words, we have $\left.\left(f v_{2}\right)\right|_{r p_{2}}=\perp$ and $r p_{1} \leq r p_{2} \leq r p_{3}$. The following lemma summarizes these observations.

Lemma 4.3.3: For all $v_{1}, v_{2}, v_{3} \in \llbracket \tau \rrbracket$ if

$$
v_{1} \rightsquigarrow r p_{1} v_{2} \mapsto_{r p_{1}} v_{3}
$$

and $\left.\left(f v_{3}\right)\right|_{r p_{3}}=\perp$ with $r p_{1} \leq r p_{3}$, then there exists $v \in \llbracket \tau \rrbracket$ such that either

$$
v_{1} \nvdash r p_{1} v \rightsquigarrow r p_{3} v_{3} \quad \text { or } \quad v_{1} \not \mapsto_{r p_{1}} v \not \longmapsto r p_{2} v_{3}
$$

for some $r p_{2}$ with $r p_{1} \leq r p_{2} \leq r p_{3}$.
We generalize the previous lemma to sequences of arbitrary length that are terminated by a sequential step. By iterated application of the previous lemma we can move this sequential step to the front of the sequence. Moving this step to the front might result in new non-sequential steps that can be moved to the front again. Finally, we get a sequence that first performs only sequential steps before it then performs only sequential steps. Furthermore, the resulting sequence is monotonic with respect to result positions. This monotonicity allows us to conclude that the pair $\left(v^{\prime}, r p^{\prime}\right)$ is in the characteristic set if the pair $\left(v_{1}, r p\right)$ is in the set as well.

Lemma 4.3.4: For all $v_{1}, v_{2}, v_{3} \in \llbracket \tau \rrbracket$ if

$$
v_{1}(\rightsquigarrow r p)^{*} v_{2} \mapsto r p v_{3}
$$

and $\left.\left(f v_{3}\right)\right|_{r p^{\prime}}=\perp$ with $r p \leq r p^{\prime}$, then there exist $v, v^{\prime} \in \llbracket \tau \rrbracket$ such that

$$
v_{1} \mapsto_{r p} v \mapsto^{*} v^{\prime}\left(\rightsquigarrow_{r p^{\prime}}\right)^{*} v_{3} .
$$

Furthermore, this sequence is montone with respect to result positions.
We use the previous lemma to show that for every pair of argument value and result position that is not in the characteristic set of a function there exists a non-sequential sequence that arrives at this pair and starts from a pair in the characteristic set.

Lemma 4.3.5: For all finite values $v \in \llbracket \tau \rrbracket$ with $\left.(f v)\right|_{r p}=\perp$ the following holds.

$$
(v, r p) \notin C_{f} \Longrightarrow \exists\left(v^{\prime}, r p\right) \in C_{f}: v^{\prime}(\rightsquigarrow r p)^{*} v
$$

The main theorem of this section shows that the characteristic set is indeed in a certain sense characteristic for a function. Functions $f$ and $g$ with $f \sqsupseteq g$ are equal if and only if $\left.(f v)\right|_{r p}=\perp$ for all pairs $(v, r p)$ of the characteristic set of $g$. Before we prove this main theorem, we prove the following lemma, which shows a connection between non-sequential refinement steps of functions that are related by $\sqsupseteq$. Note that we consider functions $f$ and $g$ with $f \sqsupseteq g$ and not $f \preceq g$. We do not need the additional constraint that the functions agree for total arguments to proof the following statements. Nevertheless, obviously these statements also hold for functions $f$ and $g$ with $f \preceq g$.

Lemma 4.3.6: Let $f, g \in \llbracket \sigma \rightarrow \tau \rrbracket$ such that $f \sqsupseteq g$. For all $v, v^{\prime} \in \llbracket \sigma \rrbracket$ if $v \rightsquigarrow_{r p}^{g} v^{\prime}$ and $\left.(f v)\right|_{r p}=\perp$, then $v \rightsquigarrow_{r p}^{f} v^{\prime}$.

Proof: Let $v, v^{\prime} \in \llbracket \tau \rrbracket$ such that $v \rightsquigarrow_{r}^{g} v^{\prime}$. We have $\left.(f v)\right|_{r p}=\perp$ and $\left.(g v)\right|_{r p}=\perp$. Let $p$ be a non-sequential position in $v$ at result position $r p$ with respect to $g$. By definition of non-sequential steps there exists a position $p$ such that $v \lessdot_{p} v^{\prime}$. Furthermore, as $p$ is a non-sequential position in $v$ at result position $r p$ with respect to $g$ there exists a value $v^{\prime \prime} \in \llbracket \tau \rrbracket$ such that $v^{\prime \prime} \sqsupset v$ and $\left.v^{\prime \prime}\right|_{p}=\perp$ and $\left.\left(g v^{\prime \prime}\right)\right|_{p} \neq \perp$. Because
$f \preceq g$, we have $f v^{\prime \prime} \sqsupseteq g v^{\prime \prime}$. Therefore, we have $\left.\left(f v^{\prime \prime}\right)\right|_{p} \neq \perp$ by monotonicity of $\left.\cdot\right|_{p}$. That is, $p$ is also a non-sequential position in $v$ at result position $r p$ with respect to $f$.

The following theorem finally shows that functions $f$ and $g$ with $f \sqsupseteq g$ are equal if we have $\left.(f v)\right|_{r p}=\perp$ for all $(v, r p) \in C_{g}$.
Theorem 4.3.2: Let $f, g \in \llbracket \tau \rightarrow \tau^{\prime} \rrbracket$ such that $f \sqsupseteq g$. The following holds.

$$
f=g \Longleftrightarrow \forall(v, r p) \in C_{g} .\left.(f v)\right|_{r p}=\perp
$$

Proof: We prove both implications separately.
$\Longrightarrow$ : Because we have $\left.(g v)\right|_{r p}=\perp$ for all $(v, r p) \in C_{g}$ and $f=g$, we get $\left.(f v)\right|_{r p}=$ $\perp$ for all $(v, r p) \in C_{g}$.
$\Longleftarrow$ : As we have observed in the introduction of this section, in the setting we are considering, if two functions agree for all finite values, they also agree for non-finite values. Therefore, we only have to consider finite values in the following.
Because $f \sqsupseteq g$, we have $f v \sqsupseteq g v$ for all $v^{\prime} \in \llbracket \tau \rrbracket$. We assume that there exists a finite value $v^{\prime} \in \llbracket \tau \rrbracket$ such that $f v^{\prime} \sqsupset g v^{\prime}$. Then, by definition of $\sqsupseteq$ there exists a result position $r p$ such that $\left.\left(f v^{\prime}\right)\right|_{r p} \neq \perp$ but $\left.\left(g v^{\prime}\right)\right|_{r p}=\perp$. Thus, we only have to show that $\left.(g v)\right|_{r p}=\perp$ implies $\left.(f v)\right|_{r p}=\perp$ for all finite values $v$.
Let $v \in \llbracket \tau \rrbracket$ be a finite value and $r p \in \operatorname{Pos}(g v)$ such that $\left.(g v)\right|_{r p}=\perp$. We have $r p \in \operatorname{Pos}(f v)$ because $f \sqsupseteq g$. If $(v, r p) \in C_{g}$, then by precondition we have $\left.(f v)\right|_{r p}=\perp$. If $(v, r p) \notin C_{g}$, by Lemma 4.3.5 there exists $\left(v^{\prime}, r p\right) \in C_{g}$ such that $v^{\prime}\left(\rightsquigarrow_{r}^{g}\right)^{*} v$. As $\left(v^{\prime}, r p\right) \in C_{g}$ and $\left.\left(g v^{\prime}\right)\right|_{r p}=\perp$, we have $\left.\left(f v^{\prime}\right)\right|_{r p}=\perp$ as well. By multiple applications of Lemma 4.3.6 and Lemma 4.3.1 we get $v^{\prime}(\leadsto f r p)^{*} v$ and, thus, $\left.(f v)\right|_{r p}=\perp$.

A reformulation of one implication of the previous theorem reveals a connection between Theorem 4.3.2 and the sufficiency of the criterion to check whether a function is minimally strict. More precisely, we observe that a less strict function is always more defined with respect to a pair from the characteristic set of the unnecessarily strict function.
Corollary 4.3.1: Let $f, g \in \llbracket \sigma \rightarrow \tau \rrbracket$. Then the following holds.

$$
f \sqsupset g \Longrightarrow \exists(v, r p) \in C_{g} .\left.(f v)\right|_{r p} \sqsupset \perp
$$

Proof: Let $f, g \in \llbracket \sigma \rightarrow \tau \rrbracket$ such that $f \sqsupset g$, that is, $f \sqsupseteq g$ and $f \neq g$. By Theorem 4.3.2 there exists $(v, r p) \in C_{g}$ such that $\left.(f v)\right|_{r p} \neq \perp$. Because we have $f \sqsupset g$, we get $\left.(f v)\right|_{r p} \sqsupset \perp$.

Finally, we show the sufficiency of the criterion, which gives us one half of the proof of Theorem 4.3.1.

Lemma 4.3.7: Let $g \in \llbracket \sigma \rightarrow \tau \rrbracket$ be sequential. If there exists a sequential function $f \in$ $\llbracket \sigma \rightarrow \tau \rrbracket$ with $f \sqsupset g$, then there exists $(v, r p) \in C_{g}$ with $\left.\left(\inf _{g} v\right)\right|_{r p} \sqsupset \perp$.

Proof: If there exists a sequential function $f$ with $f \sqsupset g$, by Corollary 4.3.1 there exists $(v, r p) \in C_{g}$ such that $\left.(f v)\right|_{r p} \sqsupset \perp$. Furthermore, for all total values $t v$ of type $\tau$ we have $f t v=g t v$, and, therefore, for all values $v$ of type $\tau$ we have $\inf _{f} v=$ $\inf _{g} v$. As we have already observed (just before Definition 4.1.3), we have $\inf _{f} v \sqsupseteq$ $f v$. Finally, we get $\left.\left(\inf _{g} v\right)\right|_{r p}=\left.\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupseteq(f v)\right|_{r p} \sqsupset \perp$.

The previous lemma also holds if we consider functions $f$ and $g$ such that $f \prec g$.

### 4.3.2 Necessity of the Criterion

In this section we show that $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$ for all $(v, r p) \in C_{f}$ is a necessary condition for $f$ being minimally strict. That is, we show that, if $f$ is minimally strict, then we have $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$ for all $(v, r p) \in C_{f}$. Actually, we prove the contraposition of this statement: if there exists $(v, r p) \in C_{f}$ with $\left.\left(\inf _{f} v\right)\right|_{r p} \neq \perp$, then $f$ is not minimally strict. To prove this statement, for a function $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ with $(v, r p) \in C_{f}$ and $\left.\left(\inf _{f} v\right)\right|_{r p} \neq \perp$, we define a function $\bar{f}_{(v, r p)}$ that is less strict than $f$. Note that the less strict function is indexed by the counter-example. That is, for every counter-example we can define a function that is less strict than $f$. Here and in the following, by counter-example we refer to a pair $(v, r p) \in C_{f}$ with $\left.\left(\inf _{f} v\right)\right|_{r p} \neq \perp$. Furthermore, note that $\left.\left(\inf _{f} v\right)\right|_{r p} \neq\left.\perp \operatorname{implies~}^{\left(\inf _{f} v\right)}\right|_{r p} \sqsupset \perp{\text { because } \inf _{f} v \sqsupseteq f v}$ and $\left.(f v)\right|_{r p}=\perp$.

If there exists $(v, r p) \in C_{f}$ with $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$, we are looking for a function $\bar{f}_{(v, r p)}$ that agrees with $\inf _{f} v$ at result position $r p$. We are looking for a function $\bar{f}_{(v, r p)}$ with

$$
\left.\left(\bar{f}_{(v, r p)} v\right)\right|_{r p}=\left.\left(\inf _{f} v\right)\right|_{r p} .
$$

As a first approach for all other arguments $v^{\prime}$ and result positions $r p^{\prime}$, the function $\bar{f}_{(v, r p)}$ may agree with $f$, that is,

$$
\left.\left(\bar{f}_{(v, r p)} v^{\prime}\right)\right|_{r p^{\prime}}=\left.\left(f v^{\prime}\right)\right|_{r p^{\prime}} .
$$

However, the following example demonstrates that we cannot use this simple definition as the resulting function is not continuous.

Example 4.3.4: Consider a function that takes a list and yields a singleton list.

```
singleton :: \([\) Bool \(] \rightarrow[\) Bool \(]\)
singleton \(x s=\)
    case \(x s\) of
            Nil \(_{\text {Bool }} \rightarrow\) Cons \(\left\langle\right.\) False, Nil \(\left._{\text {Bool }}\right\rangle\)
            Cons \(\langle y, y s\rangle \rightarrow\) case \(y\) of
                            False \(\rightarrow\) Cons \(\left\langle y\right.\), Nil \(\left._{\text {Bool }}\right\rangle\)
                            True \(\rightarrow\) Cons \(\left\langle y, N i l_{\text {Bool }}\right\rangle\)
```

This function is unnecessarily strict for two reasons. First, it checks whether the argument is the empty list or not although it yields a list with one element in all cases. Furthermore, it performs pattern matching on the first element of the argument list although it yields a singleton list in both cases. We get the following semantics for singleton.

$$
\llbracket \text { singleton』 } \rrbracket \lambda v . \begin{cases}\perp & \text { if } v=\perp \\ \text { Cons }\left\langle\text { False, Nil } l_{\text {Bool }}\right\rangle & \text { if } v=\text { Nil } l_{\text {Bool }} \\ \perp & \text { if } v=\operatorname{Cons}\left\langle\perp, v_{2}\right\rangle \\ \operatorname{Cons}\left\langle v_{1}, N i l_{\text {Bool }}\right\rangle & \text { if } v=\operatorname{Cons}\left\langle v_{1}, v_{2}\right\rangle \wedge v_{1} \neq \perp\end{cases}
$$

As we have $(\perp, \epsilon) \in C_{\llbracket \text { singleton } \rrbracket}$ as well as

$$
\left.\left(\inf _{\llbracket \text { singleton } \rrbracket} \perp\right)\right|_{\epsilon}=\operatorname{Cons}\left\langle\perp, N i l_{\text {Bool }}\right\rangle \sqsupset \perp=\llbracket \text { singleton } \rrbracket \perp,
$$

Theorem 4.3.1 would state that singleton is not minimally strict. Therefore, we are looking for a less strict function and consider the following definition of $\overline{\llbracket \text { singleton } \rrbracket}$.

$$
\overline{\llbracket \operatorname{singleton\rrbracket } v} v= \begin{cases}\text { Cons }\left\langle\perp, N i l_{B o o l}\right\rangle & \text { if } v=\perp \\ \llbracket \text { singleton } \rrbracket v & \text { otherwise }\end{cases}
$$

Obviously, $\overline{\llbracket \operatorname{singleton} \rrbracket}$ is less strict than $\llbracket \operatorname{singleton\rrbracket .~However,~this~function~is~not~}$ monotonic as we have $\perp \sqsubseteq$ Cons $\langle\perp, \perp\rangle$, but

$$
\overline{\llbracket \text { singleton } \rrbracket} \perp=\text { Cons }\left\langle\perp, N i l_{\text {Bool }}\right\rangle
$$

and

$$
\overline{\llbracket \text { singleton】 }}(\text { Cons }\langle\perp, \perp\rangle)=\perp \text {. }
$$

This example demonstrates that we may not change the definition of a function for a single argument, in this case $\perp$, as the resulting function is non-continuous. Thus, we might consider the following definition, which does not change the behavior for a single argument but for all arguments that are more defined than $\perp$.

$$
\overline{\llbracket \operatorname{singleton\rrbracket } v}= \begin{cases}\text { Cons }\left\langle\perp, N i l_{\text {Bool }}\right\rangle & \text { if } v \sqsupseteq \perp \\ \llbracket \text { singleton } \rrbracket v & \text { otherwise }\end{cases}
$$

However, in contrast to the previous definition, this time $\overline{\llbracket \operatorname{singleton} \rrbracket}$ is not less strict than $\llbracket$ singleton $\rrbracket$ as the following example demonstrates. We have

$$
\overline{\llbracket \text { singleton } \rrbracket}(\text { Cons }\langle\text { False }, \perp\rangle)=\text { Cons }\left\langle\perp, \text { Nil }_{\text {Bool }}\right\rangle
$$

and

$$
\llbracket \text { singleton } \rrbracket(\text { Cons }\langle\text { False }, \perp\rangle)=\text { Cons }\left\langle\text { False, } \text { Nil }_{\text {Bool }}\right\rangle .
$$

This example shows that we have to take over the definition of $\llbracket$ singleton $\rrbracket$ if it yields a more defined result than $\overline{\llbracket \text { singleton } \rrbracket}$.

In order to define a function that shows this behavior we use the least upper bound of the original result and the less strict result. Here and in the following, we assume that function application by juxtaposition binds stronger than the binary supremum operator $\sqcup$.

$$
\overline{\llbracket \text { singleton } \rrbracket v}= \begin{cases}\llbracket \text { singleton } \rrbracket v \sqcup \text { Cons }\left\langle\perp, N i l_{\text {Bool }}\right\rangle & \text { if } v \sqsupseteq \perp \\ \llbracket \text { singleton } \rrbracket v & \text { otherwise }\end{cases}
$$

In this case the resulting function is continuous and less strict than the original function. In the general case of an arbitrary pair $(v, r p) \in C_{f}$ with $\inf _{f} v \sqsupset \perp$ we have to spend some extra care. We have to guarantee that $\bar{f}$ is only more defined at the considered result position $r p$ and not at other result positions. Otherwise the function might become non-sequential. In the example at hand this task was easy because we considered the result position $\epsilon$.

The following definition presents a function $\bar{f}$ that is continuous, sequential, and less strict than $f$. In contrast to the previous example we use a substitution to only refine the function with respect to result position $r p$. More precisely, we replace the subterm of the result of $f$ that is undefined but should be more defined by the subterm of the supremum at the corresponding position.

Definition 4.3.4 (Less Strict Function): Let $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ and $(v, r p) \in C_{f}$ such that $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$. For all $v^{\prime} \in \llbracket \tau \rrbracket$ we define a function $\bar{f}_{(v, r p)}: \llbracket \sigma \rrbracket \rightarrow \llbracket \tau \rrbracket$ as follows.

$$
\bar{f}_{(v, r p)} v^{\prime}:= \begin{cases}\left(f v^{\prime}\right)\left[\left.\left(f v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} & \text { if } v^{\prime} \sqsupseteq v \\ f v^{\prime} & \text { otherwise }\end{cases}
$$

In the following we mostly use $\bar{f}$ instead of $\bar{f}_{(v, r p)}$.
By employing the statements of Lemma 4.2.1 and Lemma 4.2.2 we can prove that the first line is, in fact, equivalent to $f v^{\prime} \sqcup\left(f v^{\prime}\right)\left[\left.\left(\inf _{f} v\right)\right|_{r p}\right]_{r p}$.

To show that $\bar{f}$ is well-defined we have to show that the supremum $f v^{\prime} \sqcup \inf _{f} v$ always exists. Right after Definition 4.1.3 we have shown that for every value $v$ there exists a total value that is at least as defined as $v$. Therefore, for every value $v^{\prime} \in \llbracket \sigma \rrbracket$ there exists a total value $t v \in \llbracket \sigma \rrbracket \uparrow$ with $t v \sqsupseteq v^{\prime}$. Furthermore, we have $t v \sqsupseteq v$ because $v^{\prime} \sqsupseteq v$. By monotonicity we have $f t v \sqsupseteq f v^{\prime}$. By definition of inf $f$ we get $f t v \sqsupseteq \inf _{f} v$ because $\inf _{f}$ is an infimum of a set that, among other applications, contains the application $f$ tv. In a directed cpo every directed set has a least upper bound. As $\left\{f v^{\prime}, \inf _{f} v\right\}$ is a directed set - $f t v$ is an upper bound of this set - $f v^{\prime}$ and $\inf _{f} v$ have a least upper bound.

Example 4.3.5: Let us consider the functions andL from Example 2.2.1 and and from Example 4.1.2 and their characteristic sets $C_{\llbracket a n d L \rrbracket}$ and $C_{\llbracket a n d \rrbracket}$ from Example 4.3.2. We
have

$$
\begin{aligned}
\left.\left(\inf _{\llbracket a n d L \rrbracket}\langle\perp, \perp\rangle\right)\right|_{\epsilon} & =\prod\left\{\begin{array}{c}
\llbracket a n d L \rrbracket\langle\text { False, False }\rangle, \llbracket \text { andL } \rrbracket\langle\text { False, True }\rangle \\
\llbracket a n d L \rrbracket\langle\text { True,False }\rangle, \llbracket \text { andL } \rrbracket\langle\text { True, True }\rangle
\end{array}\right\} \\
& =\prod\{\text { False, False, False, True }\} \\
& =\perp
\end{aligned}
$$

and

$$
\begin{aligned}
\left.\left(\inf _{\llbracket a n d L \rrbracket}\langle\text { True, } \perp\rangle\right)\right|_{\epsilon} & =\bigcap\{\llbracket \text { andL } \rrbracket\langle\text { True, False }\rangle, \llbracket \text { andL } \rrbracket\langle\text { True, True }\rangle\} \\
& =\bigcap\{\text { False, True }\} \\
& =\perp .
\end{aligned}
$$

By Lemma 4.3.7, which is one implication of Theorem 4.3.1, this implies that all sequential functions $g$ are not less strict than $\llbracket a n d L \rrbracket$. Thus, $\llbracket a n d L \rrbracket$ is minimally strict. On the other hand we have

$$
(\langle\text { False }, \perp\rangle, \epsilon) \in C_{\llbracket a n d \rrbracket}
$$

and

$$
\begin{aligned}
\left.\left(\inf _{\llbracket a n d \rrbracket}\langle\text { False, } \perp\rangle\right)\right|_{\epsilon} & =\rceil\{\llbracket \text { andL } \rrbracket\langle\text { False, False }\rangle, \llbracket \text { andL } \rrbracket\langle\text { False, True }\rangle\} \\
& =\rceil\{\text { False,False }\} \\
& =\text { False } \\
& \sqsupset \perp
\end{aligned}
$$

as well as

$$
(\langle\perp, \text { False }\rangle, \epsilon) \in C_{\llbracket a n d \rrbracket}
$$

and

$$
\begin{aligned}
\left.\left(\inf _{\llbracket a n d \rrbracket}\langle\perp, \text { False }\rangle\right)\right|_{\epsilon} & =\rceil\{\llbracket a n d L \rrbracket\langle\text { False,False }\rangle, \llbracket \text { andL } \rrbracket\langle\text { True,False }\rangle\} \\
& =\rceil\{\text { False,False }\} \\
& =\text { False } \\
& \sqsupset \perp .
\end{aligned}
$$

The unproven implication of Theorem 4.3 .1 will show that this implies that and is not minimally strict. We will prove this missing implication by showing that the existence of $(v, r p) \in C_{\llbracket a n d \rrbracket}$ with $\left.\left(\inf _{\llbracket a n d \rrbracket} v\right)\right|_{r p} \sqsupset \perp$ implies that $\overline{\llbracket a n d \rrbracket}_{(v, r p)}$ is less strict than $\llbracket a n d \rrbracket$. Furthermore, the value of $\left.\left(\inf _{\llbracket a n d \rrbracket} v\right)\right|_{r p}$ indicates how we can improve and with respect to strictness. The implementation would be less strict if it would yield False instead of $\perp$ for the argument $\langle$ False, $\perp\rangle$ or for the argument $\langle\perp$, False $\rangle$. Note that there is no sequential function that satisfies both counter-examples.

As the pair $(\langle$ False, $\perp\rangle, \epsilon)$ as well as $(\langle\perp$, False $\rangle, \epsilon)$ are counter-examples, there are two possibilities to define functions that are less strict than and, namely,

$$
\overline{\llbracket a n d \rrbracket}{ }_{(\langle\text {False }, \perp\rangle, \epsilon)}
$$

as well as

$$
\overline{\llbracket a n d \rrbracket}_{(\langle\perp, \text { False }\rangle, \epsilon)} .
$$

As an example, the following reasoning shows that the former function yields a more defined result than $\llbracket a n d \rrbracket$ for the argument $\langle$ False,$\perp\rangle$.

$$
\begin{aligned}
& \overline{\llbracket a n d \rrbracket} \rrbracket_{(\langle\text {False }, \perp\rangle, \epsilon)}\langle\text { False }, \perp\rangle \\
& \quad=(\llbracket \text { and } \rrbracket\langle\text { False }, \perp\rangle)\left[\left.\left(\llbracket \text { and } \rrbracket\langle\text { False }, \perp\rangle \sqcup \inf _{\llbracket \text { and } \rrbracket}\langle\text { False }, \perp\rangle\right)\right|_{\epsilon}\right]_{\epsilon} \\
& \quad=\perp\left[\left.(\perp \sqcup \text { False })\right|_{\epsilon}\right]_{\epsilon} \\
&=\perp[\text { False }]_{\epsilon} \\
&=\text { False }
\end{aligned}
$$

Furthermore, it yields a result that is at least as defined as the result of $\llbracket a n d \rrbracket$ if we consider an argument that is more defined than $\langle$ False, $\perp\rangle$. In the case of $\llbracket a n d \rrbracket$ the less strict definition agrees with the infimum, for example, we have

$$
\left.\begin{array}{rl}
\llbracket \text { and } \rrbracket \\
(\langle\text { False } \perp\rangle, \epsilon)
\end{array}\langle\text { False, False }\rangle\right)
$$

Finally, if we consider a value that is less defined than or incomparable with the value $\langle$ False, $\perp\rangle$, the improved function yields the same result as the original function.

$$
\overline{\llbracket a n d \rrbracket}_{(\langle\text {False }, \perp\rangle, \epsilon)}\langle\perp, \text { False }\rangle=\llbracket \text { and } \rrbracket\langle\perp, \text { False }\rangle=\perp
$$

In summary the function $\overline{\llbracket a n d} \rrbracket_{(\langle F a l s e, \perp\rangle, c)}$ is semantically equivalent to $\llbracket a n d L \rrbracket$ while the function $\overline{\llbracket a n d \rrbracket}{ }_{(\langle\perp, F a l s e\rangle, \varepsilon)}$ is semantically equivalent to $\llbracket a n d R \rrbracket$.

As a first step towards a proof that $\bar{f}$ is well-defined, by case distinction we show that $\bar{f}$ is indeed less strict than $f$. In particular, a function $\bar{f}_{(v, r p)}$ yields a more defined result than $f$ if we consider the argument $v$.

Lemma 4.3.8: For all $\underline{f} \in \llbracket \sigma \rightarrow \tau \rrbracket$ if there exists $v \in \llbracket \sigma \rrbracket$ and $r p \in \operatorname{Pos}(f v)$ such that $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$, then $\bar{f}_{(v, r p)}$ is less strict than $f$.

The characteristic set of a function does not contain non-finite elements. For example, if we consider a function $f \in \llbracket[$ Bool $] \rightarrow[$ Bool $] \rrbracket$ the characteristic set $C_{f}$ does not contain infinite lists.

The following lemma states that the characteristic set only contains finite elements where finite is defined in Definition 2.2.1. We prove this lemma by induction over the structure of the characteristic set and by employing Lemma A.3.1.

Lemma 4.3.9: For all $(v, r p) \in C_{f}$ the value $v$ is finite.
To show that $\bar{f}$ is well-defined we have to prove that $\bar{f}$ is continuous, in other words, $\bar{f} \in \llbracket \sigma \rightarrow \tau \rrbracket$, and that $\bar{f}$ is sequential. First we show that $\bar{f}$ is a monotonic and continuous function. We prove that $\bar{f}$ is monotonic by case distinction.

Lemma 4.3.10: Let $f \in \llbracket \sigma \rightarrow \tau \rrbracket$. If there exists $(v, r p) \in C_{f}$ such that $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$, then $\bar{f}_{(v, r p)}$ is monotonic.

The following lemma shows that $\bar{f}$ is a continuous function. Note that we need the monotonicity of $\bar{f}$ to show that $\left\langle\bar{f} v_{i}\right\rangle_{i \in I}$ is a chain if we consider an arbitrary chain $\left\langle v_{i}\right\rangle_{i \in I}$. Because we consider chain-complete partial orders, this implies that the supremum

$$
\bigsqcup_{i \in I}\left(\bar{f}_{(v, r p)} v_{i}\right)
$$

exists. We prove that $\bar{f}$ is continuous by case distinction and the fact that the characteristic set contains only finite values.
Lemma 4.3.11: Let $f \in \llbracket \sigma \rightarrow \tau \rrbracket$. If there exists $(v, r p) \in C_{f}$ such that $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$, then $\bar{f}_{(v, r p)}$ is continuous.

To prove that $\bar{f}$ is sequential we have to prove a property about the characteristic set $C_{f}$. If we consider an arbitrary pair $(v, r p)$, which is not an element of the characteristic set, for the definition of $\bar{f}_{(v, r p)}$ the resulting function might not be sequential. Therefore, the sequentiality of $\bar{f}$ has to rely on some property of the characteristic set. The following example provides a function $\bar{f}_{(v, r p)}$ that is not sequential if we use an arbitrary pair $(v, r p)$ that is not an element of the characteristic set of $f$.

Example 4.3.6: For example, consider the function andL again. We have

$$
\left.\left(\inf _{\llbracket a n d L \rrbracket}\langle\perp, \text { False }\rangle\right)\right|_{\epsilon}=\text { False } \sqsupset \perp .
$$

Although the pair $(\langle\perp, F a l s e\rangle, \epsilon)$ is not in the characteristic set of andL, we consider the function $\overline{\llbracket a n d L \rrbracket}{ }_{(\langle\perp, F a l s e\rangle, \epsilon)}$ in the following. We have

$$
\overline{\llbracket a n d L \rrbracket}_{(\langle\perp, \text { False }\rangle, \epsilon)}\langle\text { False }, \perp\rangle=\llbracket \text { andL } \rrbracket\langle\perp, \text { False }\rangle=\text { False }
$$

as well as

$$
\begin{aligned}
& \overline{\llbracket a n d L \rrbracket} \rrbracket_{(\langle\perp, \text { False }\rangle, \epsilon)}\langle\perp, \text { False }\rangle \\
& =(\llbracket \text { andL } \rrbracket\langle\perp, \text { False }\rangle)\left[\left.\left(\llbracket a n d L \rrbracket\langle\perp, \text { False }\rangle \sqcup \inf _{\llbracket a n d L \rrbracket}\langle\perp, \text { False }\rangle\right)\right|_{\epsilon}\right]_{\epsilon} \\
& =\perp\left[\left.(\perp \sqcup \text { False })\right|_{\epsilon}\right]_{\epsilon} \\
& =\text { False. }
\end{aligned}
$$

Therefore, the function considered above is not sequential as there exists no sequential position in $\langle\perp, \perp\rangle$ at position $\epsilon$. This example demonstrates that we may not
consider arbitrary pairs of values and result positions for the definition of $\bar{f}$ but only pairs that belong to $C_{f}$.

Let us assume that we want to prove that $\bar{f}$ is sequential. That is, for every partial value $p v$ we have to provide a sequential position in $p v$. More precisely, we have to provide some position $p$ such that for all $p v^{\prime}$ with $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p}=\perp$ we have $\left.\left(\bar{f} p v^{\prime}\right)\right|_{r p}=\perp$. Now consider that there exists some partial value $p v^{\prime}$ with $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p}=\perp$ such that $p v^{\prime} \sqsupseteq v$ while we have $p v \nexists v$ where $(v, r p) \in C_{f}$ is the counter-example used to define $\bar{f}$. By definition of $\bar{f}$ we have

$$
\bar{f} p v=f p v
$$

but

$$
\bar{f} p v^{\prime}=\left(f p v^{\prime}\right)\left[\left.\left(f p v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} .
$$

Furthermore, we have

$$
\begin{aligned}
\left.\left(\bar{f} p v^{\prime}\right)\right|_{r p} & =\left.\left(\left(f p v^{\prime}\right)\left[\left.\left(f p v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p}\right)\right|_{r p} & & \\
& =\left.\left(f p v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p} & & \text { Lemma 4.2.2 } \\
& =\left.\left.\left(f p v^{\prime}\right)\right|_{r p} \sqcup\left(\inf _{f} v\right)\right|_{r p} & & \text { continuity of }\left.\cdot\right|_{r p}
\end{aligned}
$$

As $(v, r p)$ is a counter-example, we have $\left.\left(\inf _{f} v\right)\right|_{r p} \neq \perp$, and, therefore, $\left.\left(\bar{f} p v^{\prime}\right)\right|_{r p} \neq$ $\perp$. That is, although there exists a sequential position in $p v$ with respect to $f$ this position is not sequential with respect to $\bar{f}$. Thus, in this case we have to prove that there exists another sequential position. More precisely, we need a sequential position such that for all $p v^{\prime}$ with $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p}=\perp$ we have $p v^{\prime} \nexists v$ to avoid the problem described above.

In Example 4.3.6 the role of $p v$ is taken over by $\langle\perp, \perp\rangle$ and the role of $p v^{\prime}$ is taken over by $\langle\perp$, False $\rangle$. Furthermore, we have $v=\langle\perp$, False $\rangle$, and, therefore, $p v \nsupseteq v$ but $p v^{\prime} \sqsupseteq v$. The definition of $\overline{\llbracket a n d L \rrbracket}$ "destroys" the sequential position in $\langle\perp, \perp\rangle$ with respect to andL, namely, position 1 . But, as we did consider a pair $(v, r p)$ that is not in the characteristic there exists no other sequential position in $\langle\perp, \perp\rangle$, and, therefore, $\overline{\llbracket a n d L \rrbracket}$ is not sequential. In the following we illustrate why there always exists a sequential position with the desired property if we only consider counter-examples from the characteristic set.

The following lemma proves the property of the elements of the characteristic set of a function we have just illustrated. In Lemma 4.3.13 we employ this property to prove that $\bar{f}$ is a sequential function if $f$ is a sequential function. We prove the following lemma by employing Lemma A.3.1.

Lemma 4.3.12: Let $p$ be a sequential position of $f$ in $p v$ at position $r p$ and $(v, r p) \in C_{f}$ with $v \nsubseteq p v$. Then there exists a sequential position $p^{\prime}$ of $f$ in $p v$ at position $r p$ such that for all $p v^{\prime} \in \llbracket \tau \rrbracket$ with $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p^{\prime}}=\perp$ we have $p v^{\prime} \nsupseteq v$.

With the help of the previous lemma we are ready to prove that $\bar{f}_{(v, r p)}$ is sequential if there exists $(v, r p) \in C_{f}$ such that $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$.

Lemma 4.3.13: Let $f \in \llbracket \sigma \rightarrow \tau \rrbracket$. If there exists $(v, r p) \in C_{f}$ such that $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$, then $\bar{f}_{(v, r p)}$ is a sequential function.

The following lemma finally proves that the presented criterion is necessary for the existence of a less strict, sequential function and constitutes the missing halve to the proof of Theorem 4.3.1.

Lemma 4.3.14: Let $f \in \llbracket \sigma \rightarrow \tau \rrbracket$ be a sequential function. If there exists $(v, r p) \in C_{f}$ such that $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$, then there exists a sequential function $g \in \llbracket \sigma \rightarrow \tau \rrbracket$ with $g \prec f$.

Proof: As there exists $(v, r p) \in C_{f}$ with $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$, we can define $\bar{f}_{(v, r p)}$ and we have $\bar{f} \prec f$ by Lemma 4.3.8. By Lemma 4.3.10 and Lemma 4.3 .11 we have $\bar{f} \in$ $\llbracket \sigma \rightarrow \tau \rrbracket$ and by Lemma 4.3.13 $\bar{f}$ is sequential.

## 5 Implementation of Sloth

In this chapter we present the implementation of Sloth. More precisely, we do not present the implementation by means of source code but the basic approach in an abstract manner and by means of examples.

Sloth can be used to check whether a function is unnecessarily strict. For example, consider the Haskell implementation of the function and from Example 4.1.2.

```
and :: Bool }->\mathrm{ Bool }->\mathrm{ Bool
and False False = False
and False True = False
and True False = False
and True True = True
```

Sloth provides a function that is called strictCheck to check whether a function is minimally strict for inputs up to a specific size. The size of a term is the number of constructors in the term. For example, the following application checks whether and is minimally strict for all pairs of Boolean values up to size five.

```
> strictCheck and 5
2: \False \perp -> False
Finished 7 tests.
```

Sloth reports one counter-example that states that the evaluation of the application and False $\perp$ yields $\perp$ while there exists a less strict implementation that yields False instead. The highlighting on the right-hand side of $->$ denotes that there exists an implementation that yields the highlighted value while the current implementation yields $\perp$ instead of the highlighted value.

The remainder of this chapter shows how Sloth enumerates test cases and checks whether a test case is a counter-example. Here, a test case is a value of the argument type of the function. A counter-example is a test case together with the current result of the function for this argument and a proposed, more defined, result. In contrast to Chapter 4, Sloth does not use an uncurried notation but the Haskell style curried notation for test cases. In the following, we use both notations synonymously. For example, the counter-example for and consists of the test case $\langle$ False,$\perp\rangle$, the current result $\perp$, and the proposed result False. Furthermore, in this chapter we present functions in Haskell syntax and abstain from presenting their counterparts in our simple functional language. Besides, we use Haskell-style list constructors [] and (:) instead of Nil and Cons $\langle\cdot, \cdot\rangle$ to emphasize the connection between the output of Sloth and our considerations on the level of semantics.

We refer to the symbol $\perp$ as error because this symbol, in fact, denotes a run-time error in this context. Nevertheless, it could also denote a non-terminating computation as the considered functions behave equally for run-time errors and non-terminating computations. This observation only holds if we disregard certain kinds
of exception handling mechanisms as a function otherwise could behave differently for a run-time error than for a non-terminating computation.

As Theorem 4.3.1 proves, to check whether a function $f$ is minimally strict, we have to check whether we have $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$ for all pairs $(v, r p)$ of the characteristic set of $f$. A pair $(v, r p)$ consists of a value $v$ of the argument type of $f$ and a position $r p$ that references a sub-value of the value $f v$. The vertical bar followed by the position $r p$ denotes the projection to position $r p$. For now we consider $\inf _{f} v$ as some abstract value and explain its definition in more detail later.

Intuitively, a pair $(v, r p)$ is an element of the characteristic set of a function $f$, if the function may yield a more defined result for the argument $v$ at result position $r p$ without contradicting sequentiality. That is, if we consider a function $f^{\prime}$ that is equivalent to $f$ but yields a more defined result for the argument $v$ at result position $r p$, then $f^{\prime}$ is still sequential. Besides, for all elements $(v, r p)$ of the characteristic set the current implementation of $f$ yields an error for the argument $v$ at result position $r p$. Thus, yielding a more defined result for an element $(v, r p)$ means yielding a result that is not an error. In summary, the elements of the characteristic set are candidates for making a function less strict. In Section 5.1 we illustrate how Sloth enumerates the elements of the characteristic set.

To check whether there exists a sequential function that is less strict than a function $f$, we check whether we have $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$ for all pairs $(v, r p)$ of the characteristic set. Intuitively, the value $\left.\left(\inf _{f} v\right)\right|_{r p}$ is not an error if $f$ yields similar results at result position $r p$ for arguments that are more defined than $v$. The value $\left.\left(\inf _{f} v\right)\right|_{r p}$ represents this similarity. As the function behaves similar for all more defined inputs, we can define a function that yields this similarity instead of an error for the argument $v$ at result position $r p$. In other words, there exists a function that yields more parts of the result of $f$ by inspecting the same part of its argument. In Section 5.2 we show how Sloth checks whether we have $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$.

### 5.1 Enumerating Test Cases

So, how can we enumerate the elements of the characteristic set of a function $f$ ? If a pair $(v, r p)$ is an element of the characteristic set, by Definition 4.3.3 the value $v$ is either the least element of the argument type of $f$ or there exists another pair $\left(v^{\prime}, r p^{\prime}\right)$ in the characteristic set such that $v^{\prime} \lessdot_{p} v$ for some position $p$. For now we ignore the additional requirements on $r p$ and $r p^{\prime}$ that are set up by the definition of the characteristic set.

Two values $v$ and $v^{\prime}$ are related by $v \lessdot_{p} v^{\prime}$ if we get $v^{\prime}$ from $v$ by replacing an occurrence of an error in $v$ at position $p$ by a constructor applied to errors only. Let us consider the relation $\lessdot_{p}$ for values of type Bool $\times$ Bool. We can represent this relation $\lessdot_{p}$ by a tree as shown in Figure 5.1.1. The root note of the tree is the least element of the corresponding type. Each value is followed by an argument position that contains an error, in other words, if a value $v$ is followed by a position $p$, we have $\left.v\right|_{p}=\perp$. Furthermore, if two subsequent values $v$ and $v^{\prime}$ have an intermediate argument position $p$, we have $v \lessdot_{p} v^{\prime}$. For example, in Figure 5.1.1 between the


Figure 5.1.1: Tree representing the Relation $\lessdot_{p}$ for Pairs of Boolean Values.
values $\langle$ True, $\perp\rangle$ and $\langle$ True, True $\rangle$ there is a node labeled with 2 because we have $\langle$ True, $\perp\rangle \lessdot_{2}\langle$ True, True $\rangle$.
If we omit the position nodes, the tree in Figure 5.1.1 resembles the Hasse diagram for the ordering $\sqsubseteq$ for values of type Bool $\times$ Bool. The position nodes are actually redundant as we can always deduce the position from the values of the preceding and the subsequent value. Nonetheless, we keep these redundant nodes as it makes this information explicit. Additionally, while the corresponding Hasse diagram is a directed acyclic graph, for simplicity, here as well as in the implementation of Sloth we use trees. As future work, we might check whether the performance of the implementation improves when we switch from trees to directed acyclic graphs. However, in most cases Sloth only generates small parts of these trees because of lazy evaluation.

If we consider a type that has infinitely many values, the corresponding tree might have infinitely many nodes as well. Therefore, we cut the tree at the depth that corresponds to the number of constructors passed as second argument to strictCheck. If a value $v$ precedes a value $v^{\prime}$ in the tree that represents $\lessdot$, then $v$ has exactly one constructor less than $v^{\prime}$.

To calculate the characteristic set - as we will later see - and to calculate the value $\left.\left(\inf _{f} v\right)\right|_{r p}$, we need the results of applications of $f$ to values that are more defined than some value $v$. In Figure 5.1.1 the subtree that is rooted at some value $v$ contains all values that are more defined than $v$. Therefore, we add the results of applying the considered function to each node of the tree that represents the relation $\lessdot$

As an example let us consider the function andL, which is the Haskell implementation of the function with the same name from Example 2.2.1.


Figure 5.1.2: Tree representing the Relation $\lessdot_{p}$ for Pairs of Boolean Values with Results of Applying andL to these Values.

```
andL \(::\) Bool \(\rightarrow\) Bool \(\rightarrow\) Bool
andL False \({ }_{-}=\)False
andL True \(x=x\)
```

Instead of a node $\langle$ False, $\perp\rangle$, with respect to the Boolean conjunction andL we consider a node of the form $\lambda$ False $\perp \rightarrow$ False. The value on the left-hand side of $\rightarrow$ is the argument and the value on the right-hand side is the result of applying the considered function to this argument. We refer to this kind of node as mapping node.

Figure 5.1.2 shows the result of applying andL to every node of the tree in Figure 5.1.1. To gain additional information we label errors in the argument value with their position in the argument by using positions as presented in Definition 4.2.2. Intuitively, an error like $\perp_{1}$ is represented by a run-time error that carries the position 1 as error message. By employing exception handling mechanisms, we regain these error messages from the result of an application as it is modeled by the exception semantics, presented in Figure 4.2.6. For example, as andL performs pattern matching on its first argument, the resulting error has the same label as the one in the first argument, namely, position 1. Using this mechanism, for every result position that contains an error we can determine the corresponding demanded position. For example, as we have $\llbracket a n d L \rrbracket\left\langle\perp_{1}, \perp_{2}\right\rangle=\perp_{1}$, by Definition 4.2 . 8 position 1 is the demanded position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ with respect to andL. The demanded position is the position in the argument that is evaluated to head normal form if we evaluate the value at a certain result position to head normal form. Note that the definition of demanded positions requires that every label of an error is unique. This requirement is satisfied as each error is labeled with its position in the argument value.

The tree in Figure 5.1.2 does not contain the nodes $\lambda$ False False $\rightarrow$ False as well as $\lambda$ False True $\rightarrow$ False. We omit successors of mapping nodes whose result value is total. By monotonicity of the considered functions we know that all successors of a


Figure 5.1.3: The Refinement Tree for andL, the Highlighted Part represents the Characteristic Set for andL.
node like this will yield the same result. For example, as we have $\llbracket a n d L \rrbracket\langle$ False, $\perp\rangle=$ False we get $\llbracket a n d L \rrbracket\langle$ False, False $\rangle=$ False as well as $\llbracket a n d L \rrbracket\langle$ False, True $\rangle=$ False .

Our final goal is to calculate the characteristic set of andL. As an element of the characteristic set is a pair consisting of a value and a result position, we have to add result positions to the tree in Figure 5.1.2. Intuitively, to calculate the characteristic set, for every value $v$ and every result position $r p$ such that $\left.(f v)\right|_{r p}=\perp$, we have to identify all corresponding sequential positions. Therefore, we define a tree in which every value $v$ is followed by all eligible result positions. Each result position is itself followed by all possible sequential positions with respect to the preceding result position and the preceding value. This way we define a tree that represents a superset of the characteristic set. We refer to this tree as refinement tree. To calculate the characteristic set we prune the refinement tree. Figure 5.1.3 presents the refinement tree for andL. In the following, we explain the three kinds of nodes of a refinement tree in detail.
mapping node A mapping node is labeled with a value of the argument type and the result of the function for this argument. For example, the root node in Figure 5.1.3, labeled with $\lambda \perp_{1} \perp_{2} \rightarrow \perp_{1}$, is a mapping node. To gain additional information we label errors in the argument value with their position in the argument. Using these labels, at each result position we can easily identify the corresponding demanded position.


Figure 5.1.4: A Path of a Refinement Tree.
result position node For every position of an error in the result value of a mapping node, the node has a successor that is labeled with this position. We refer to this kind of nodes as result position nodes. For example, the mapping node labeled with $\lambda \perp_{1} \perp_{2} \rightarrow \perp_{1}$ has only one child, labeled with $\epsilon$, because $\epsilon$ is the only position of an error in the result value of this mapping node, namely, $\perp_{1}$. Result position nodes of a refinement tree are ordered by the position ordering $\leq$. If a result position node labeled with $r p$ is an ancestor of a result position node labeled with $r p^{\prime}$, then $r p$ is a prefix of $r p^{\prime}$. We may omit other result positions by the definition of the characteristic set as we will observe soon.
argument position node For every position of an error in the argument value of a mapping node, the corresponding result position nodes have a successor that is labeled with the position of the error in the argument value. We refer to this kind of nodes as argument position nodes. For example, the successors of the topmost result position node, labeled with $\epsilon$, are labeled with 1 and 2 because 1 and 2 are the positions of errors in the argument value of the preceding mapping node, namely, $\lambda \perp_{1} \perp_{2} \rightarrow \perp_{1}$. In Figure 5.1 .3 we additionally mark argument position nodes that correspond to demanded positions with respect to the preceding result position node.

In summary, on a path between two mapping nodes there is always a result position node and an argument position node as illustrated in Figure 5.1.4. Here $v_{1}$ and $v_{2}$ are argument values and $r_{1}$ and $r_{2}$ are result values. Furthermore, $r p$ is a result position such that $\left.r_{1}\right|_{r p}=\perp, r p^{\prime}$ is a result position such that $\left.r_{2}\right|_{r p^{\prime}}=\perp$, and $p$ is an argument position such that $\left.v_{1}\right|_{p}=\perp$.

By definition of the characteristic set, the root node of a refinement tree is an element of the characteristic set. More precisely, all pairs that contain the argument of the root mapping node and the result position of a subsequent result position node are elements of the characteristic set. In the case of the refinement tree for andL the pair $(\langle\perp, \perp\rangle, \epsilon)$ is an element of the characteristic set, because $\epsilon$ is the result position of a result position node that succeeds the mapping node $\lambda \perp_{1} \perp_{2} \rightarrow \perp_{1}$.

When we consider a path as shown in Figure 5.1.4, if the pair $\left(v_{1}, r p\right)$ is in the characteristic set, then the pair $\left(v_{2}, r p^{\prime}\right)$ is in the characteristic set as well if the step from $v_{1}$ to $v_{2}$ is sequential. Furthermore, $r p^{\prime}$ has to be a prefix of $r p$ and we need
$\left.\left(f v_{2}\right)\right|_{r p^{\prime}}=\perp$. The last two requirements are satisfied by the definition of the refinement tree. Therefore, if the pair $\left(v_{1}, r p\right)$ is an element of the characteristic set, we only have to check whether the step from $v_{1}$ to $v_{2}$ is sequential to observe that $\left(v_{2}, r p^{\prime}\right)$ is also an element of the characteristic set. By definition of the refinement tree we have $v_{1} \lessdot_{p} v_{2}$ and $p$ is either a sequential or a non-sequential position. That is, all argument position nodes either correspond to a sequential or a non-sequential position by the definition of the refinement tree. In the rest of this section we illustrate how we identify whether a position is sequential or non-sequential.

To determine additional elements of the characteristic set in Figure 5.1.3 besides the root node, we have to identify sequential positions in $\langle\perp, \perp\rangle$ at result position $\epsilon$. Position 1 and position 2 are possible sequential positions in $\langle\perp, \perp\rangle$ at result position $\epsilon$. Lemma 4.2.5 states that all demanded positions are sequential positions. Thus, we can identify one sequential position easily. Position 1 is a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ because it is a demanded position in $\langle\perp, \perp\rangle$ at result position $\epsilon$. This way we can quite efficiently identify one sequential position for every mapping node and every subsequent result position node.

As we have observed in Example 4.2.9, there are sequential positions that are not demanded positions. Therefore, we have to check whether any of the other possible sequential positions is indeed a sequential position. For example, position 2 might be a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$. To eliminate non-sequential positions, Sloth searches for witnesses that prove that an argument position is a non-sequential position. For example, let us consider the possible sequential positions in $\langle\perp, \perp\rangle$ at result position $\epsilon$, namely, position 2. The mapping node labeled with $\lambda$ False $\perp_{2} \rightarrow$ False proves that position 2 is not a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ as we have already observed in Example 4.2.5. To show that $p$ is a non-sequential position in $v$ at result position $r p$, by definition of non-sequential positions we have to find a value $v^{\prime}$ such that $v^{\prime} \sqsupseteq v$ and $\left.v^{\prime}\right|_{p}=\perp$ but $\left.\left(f v^{\prime}\right)\right|_{r p} \neq \perp$. In the particular case at hand we have $\langle$ False, $\perp\rangle \sqsupseteq\langle\perp, \perp\rangle$ and

$$
\left.\langle\text { False }, \perp\rangle\right|_{2}=\perp \quad \text { but }\left.\quad(\llbracket \text { and } L \rrbracket\langle\text { False }, \perp\rangle)\right|_{\epsilon} \neq \perp .
$$

Therefore, position 2 is not a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ with respect to andL.

To check whether a position $p$ is a non-sequential position in a value $v$ at position $r p$, Sloth searches the subtree for mapping nodes of the form $\lambda v^{\prime} \rightarrow r^{\prime}$ such that

$$
\left.v^{\prime}\right|_{p}=\perp \quad \text { and }\left.\quad r^{\prime}\right|_{r p} \neq \perp
$$

A mapping node with this property shows that position $p$ is not a sequential position in $v$ at result position $r p$. Note that we have $v^{\prime} \sqsupseteq v$ as $v^{\prime}$ is an element of the subtree that is rooted at the mapping node, whose argument is $v$. Furthermore, note that a single witness might prove that a couple of argument positions are non-sequential.

We do not have to randomly search the tree for witnesses. Instead we can employ the information that is gained by identifying demanded positions. Because position 1 is the demanded position and, therefore, a sequential position in $\langle\perp, \perp\rangle$, at result position $\epsilon$ we have $\llbracket a n d L \rrbracket\langle\perp, v\rangle=\perp$ for all values $v$ of type Bool. Therefore, we
will not discover a witness (arguments $v_{1}$ and $v_{2}$ such that $\llbracket a n d L \rrbracket\left\langle v_{1}, v_{2}\right\rangle \neq \perp$ ) as long as the first argument is $\perp$. When we search a tree for witnesses, at result position nodes we only consider subtrees with respect to argument position nodes whose positions are demanded positions. For example, we consider the mapping node $\lambda \perp_{1} \perp_{2} \rightarrow \perp_{1}$. Because the argument position node labeled 1 is marked as demanded position, we only search the subtree that is rooted at this argument position node for witnesses.

As position 2 is not a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$, we only keep the highlighted part of the refinement tree in Figure 5.1.3. In this case the resulting tree is already the characteristic set. For each result position node the tree contains only a single argument position node, which is the demanded position in each case.

If we consider a function whose argument type has infinitely many values, checking whether a position is non-sequential cannot be done this way. We might simply miss a witness because we do not consider arguments of sufficiently large sizes. For now, we ignore this case but consider it in Section 5.4.

### 5.2 Checking Test Cases

After we have identified the tree that represents the characteristic set, we have to check whether we have $\left.\left(\inf _{f} v\right)\right|_{r p}=\perp$ for all argument values $v$ of mapping nodes and all subsequent result positions $r p$. All pairs $(v, r p)$ with $\left.\left(\inf _{f} v\right)\right|_{r p} \neq \perp$ are counter-examples. As we have

$$
\inf _{f} v=\prod\{f t v \mid t v \in \llbracket \tau \rrbracket \uparrow, v \sqsubseteq t v\},
$$

for every value $v$ in the tree we have to compute the set $\{f t v \mid t v \in \llbracket \tau \rrbracket \uparrow, v \sqsubseteq t v\}$. For example, to check whether mapping node $\lambda \perp_{1} \perp_{2} \rightarrow \perp_{1}$ together with result position $\epsilon$ is a counter-example we search the corresponding subtree for mapping nodes whose arguments are total values. In this particular case, we apply andL to $\langle$ False, $\perp\rangle,\langle$ True, False $\rangle$, and $\langle$ False, False $\rangle$ and collect the values False, True, and False. This way we observe that we have

$$
\left.\left(\inf _{\llbracket a n d L \rrbracket}\langle\perp, \perp\rangle\right)\right|_{\epsilon}=\left.\left(\prod\{\text { False,False, True }\}\right)\right|_{\epsilon}=\left.\perp\right|_{\epsilon}=\perp,
$$

and, hence, the partial value $\langle\perp, \perp\rangle$ together with result position $\epsilon$ is not a coun-ter-example. When we search for total values, we, again, only consider demanded positions. Note that the first argument, namely, $\langle$ False, $\perp\rangle$, is not a total value. A mapping node whose result value is a total value does not have any successors.

In the same way as we have observed that $(\langle\perp, \perp\rangle, \epsilon)$ is not a counter-example, we observe that all elements of the characteristic set of andL are no counter-examples, thus, by Theorem 4.3.1 andL is minimally strict.

The argument type of andL has only finitely many values. If we consider functions whose argument type has infinitely many values, we are faced with additional problems. In the following we present the approach that is used to address these
problems. We illustrate that Sloth distinguishes the following kinds of counter-examples. Sloth uses the same highlighting as the words "potential" and "definite" to indicate the corresponding kind of counter-example.
definite counter-example A definite counter-example stays a counter-example no matter what size we use for the test cases.
potential counter-example A potential counter-example might be no counter-example if we consider inputs of larger sizes. That is, there might be a size greater than the current size such that Sloth does not report the potential counter-example anymore.

In the following we explain the difference between the two kinds of counter-examples in detail. If the argument type $\tau$ of a function $f$ has infinitely many elements, the set

$$
\text { Total } v:=\{f t v \mid t v \in \llbracket \tau \rrbracket \uparrow, v \sqsubseteq t v\}
$$

might be infinite for some value $v$ of type $\tau$. To check whether an element of the characteristic set is a counter-example, we have to calculate the greatest lower bound of the set Total v. If we approximate this set, that is, we calculate a set $T^{\prime}$ that is a proper subset of Total $v$, we get $\Pi T^{\prime} \sqsupseteq \Pi$ Total $v$ because the operator $\Pi$ is monotonically decreasing. Let us consider that we have discovered a counter-example, namely, a pair $(v, r p)$ with $\left.\left(\sqcap T^{\prime}\right)\right|_{r p} \sqsupset \perp$. If we consider an approximation $T^{\prime \prime}$ that is more precise than the approximation $T^{\prime}$, we might have

$$
\left\lceil T^{\prime} \sqsupset \prod T^{\prime \prime} \quad \text { and }\left.\quad\left(\prod T^{\prime}\right)\right|_{r p}=\perp .\right.
$$

Thus, with respect to the more precise approximation the pair $(v, r p)$ is not a coun-ter-example anymore.

The tool presented by Chitil (2006) classifies all arguments $v$ with $\Pi T^{\prime} \sqsupset f v$ as unnecessarily strict. In contrast, we try to identify counter-examples that stay coun-ter-examples even if we consider an arbitrarily precise approximation. Consider the following definition.

```
potential :: \([\) Bool \(] \rightarrow\) Bool
potential [] \(=\) True
potential \((-: x s)=\) potential \(x s\)
```

Obviously, this function is unnecessarily strict. It performs pattern matching although it yields True for all total inputs. Sloth reports the following counter-examples if we check potential for Boolean lists up to size three.

```
> strictCheck potential 3
1: \\perp -> True
3: \( }\perp:\perp) -> Tru
Finished 7 tests.
```

In contrast to the counter-examples for intersperse from Section 3.2, these counter-examples are potential counter-examples. That is, there might be a size greater than
three such that Sloth does not report some of the counter-examples it reports for size three. One way to confirm a potential counter-example is to increase the size. However, Sloth still reports only potential counter-examples for the function potential if we consider arguments up to size four. The counter-examples even stay potential no matter what size we use.

If we take a closer look at the test cases, we might wonder why the largest coun-ter-example has only one constructor although we have checked potential for lists up to size three, that is, lists with at most three constructors. When we check potential for lists up to size three, using verboseCheck, we get the following result. In contrast to strictCheck, the function verboseCheck additionally reports successful test cases.

```
> verboseCheck potential 3
1: \\perp -> True
2: \[] -> True
3: \( }\perp:\perp) -> True
4: \(\perp:[]) -> True
5: \( }\perp:\perp:\perp) -> ?
6: \(\perp:\perp:[]) -> True
7: \( }\perp:\perp:\perp:\perp) -> ?
Finished 7 tests.
```

As the output illustrates that current Sloth implementation checks lists up to size three but does show all results when we use strictCheck.

If the size of the approximation $T^{\prime}$ of the set Total $v$ is very small, the result is quite likely to be wrong. For example, if we consider a set $T^{\prime}$ with a single element to approximate the set Total $v$, then we most certainly have $\left.\left(\Pi T^{\prime}\right)\right|_{r p} \sqsupset \perp$. Therefore, Sloth only considers approximations of the set Total $v$ with at least three elements. The question marks on the right-hand sides of -> state that the number of elements of the approximation is too small in these cases. By using a configuration we can specify the minimum number of elements for the approximation of the set Total $v$. Although there are probably more useful heuristics than demanding a fixed number of elements, we have not explored other heuristics.

Let us go back to the problem of identifying definite counter-examples. Consider the following variation of the function potential.

```
potential' :: [Bool] \(\rightarrow\) Bool
potential' ( \(\left.\left.-:_{-}:-\right)_{-}\right)=\)False
potential \({ }^{\prime} \quad=\quad\) True
```

This function is semantically equivalent to potential with respect to arguments up to size two. Therefore, if we only consider total values up to size two, potential' is unnecessarily strict. The function is supposed to yield the value True without inspecting its argument as potential' yields True for all total arguments up to size two. However, if we consider the total value [False,False], then we observe that potential ${ }^{\prime}$ is not unnecessarily strict as we have $\llbracket$ potential ${ }^{\prime} \rrbracket[$ False, False $]=$ False. That is, obviously, in general, we cannot decide whether a test case is a counter-example by only considering a finite number of total values.

We can manually verify a potential counter-example. A potential counter-example is definitely a counter-example if all more defined total inputs lead to results that are at least as defined as the recommended result. For example, consider the first counter-example for potential. For all total inputs that are more defined than $\perp$ the function potential yields True, which is as defined as the recommended result True. Therefore, the first counter-example is definitely a counter-example. Note that in most cases it is easy to verify a potential counter-example as we only have to consider the behavior of a function with respect to total values.

Although some counter-examples are only potential counter-examples, there are cases in which a finite approximation is sufficient to identify a definite counter-example. For example, consider the following function.

```
definite:: [Bool] }->[\mathrm{ Bool ]
definite [] = [True]
definite (_:xs) = True:definite xs
```

Both counter-examples for definite are definite counter-examples as it is indicated by the highlighting.

```
> strictCheck definite 2
1: \\perp -> True: }
3: \( }\perp:\perp) -> True:True: 
Finished 5 tests.
```

Sloth employs monotonicity to identify definite counter-examples. Let us consider the first counter-example. Sloth observes that we have
【definite』 ] [] = True: []
as well as

$$
\llbracket \text { definite } \rrbracket(\perp: \perp)=\text { True }: \perp .
$$

Informally, these equations state that, no matter whether the argument of definite is an empty or a non-empty list, the result is a list with at least one element whose first element is True. We can formally justify this statement as follows. As definite is monotonic, we have

$$
\llbracket d e f i n i t e \rrbracket v \sqsupseteq \text { True }: \perp
$$

for all Boolean lists $v$ such that $v \sqsupseteq \perp: \perp$. To check whether definite is unnecessarily strict for the argument $\perp$ at result position $\epsilon$ we have to show that the following inequality holds.

$$
\left.\left(\inf _{\llbracket d e f i n i t e \rrbracket} \perp\right)\right|_{\epsilon} \sqsupset \perp .
$$

We can show this strict inequality by the following estimate.

$$
\begin{aligned}
\left.\left(\inf _{\llbracket d e f i n i t e \rrbracket} \perp\right)\right|_{\epsilon} & =\prod\{\llbracket \text { definite } \rrbracket t v \mid t v \in \llbracket[\text { Bool }] \rrbracket \uparrow, \perp \sqsubseteq t v\} \\
& =\llbracket \text { definite } \rrbracket[ \rceil \sqcap\rceil\{\llbracket \text { definite } \rrbracket t v \mid t v \in \llbracket[\text { Bool }] \rrbracket \uparrow, \perp: \perp \sqsubseteq t v\}
\end{aligned}
$$

$$
\begin{aligned}
& =\text { True }:[] \sqcap \bigcap\{\llbracket \text { definite } \rrbracket t v \mid t v \in \llbracket[\text { Bool }] \rrbracket \uparrow, \perp: \perp \sqsubseteq t v\} \\
& \sqsupseteq \text { True }:[] \sqcap \llbracket \text { definite } \rrbracket(\perp: \perp) \\
& =\text { True }:[] \sqcap \text { True }: \perp \\
& =\text { True }: \perp
\end{aligned}
$$

Hence，we can observe that definite is unnecessarily strict for the input $\perp$ by only considering two applications．As a side effect this approach allows us to evaluate only a small part of the refinement tree to check whether an application is unneces－ sarily strict．More precisely，to observe that the application $\llbracket d e f i n i t e \rrbracket ~ \perp$ is unneces－ sarily strict we even do not have to search the tree for applications of definite to total values．Instead，we only have to evaluate a single level of the tree as the applications【definite】［］and 【definite】 $(\perp: \perp)$ already provide sufficient information．

In the case of definite the estimate is actually an equality．However，there are cases in which we get a strict inequality instead．As a consequence，the counter－examples reported by Sloth may not agree with the behavior of a minimally strict implementa－ tion．For example，consider the following definition，which is a variation of definite．

```
definite \(^{\prime}::[\) Bool \(] \rightarrow[\) Bool \(]\)
definite \(^{\prime}[]=[\) True,True \(]\)
definite \((-: x s)=\)
    True : case \(x s\) of
    [] \(\rightarrow\) [True]
    \(-:\) ys \(\rightarrow\) True: definite' ys
```

If we check this function，Sloth reports the following counter－examples．

```
> strictCheck definite' 2
1: \\perp -> True:\perp
3: \( }\perp:\perp) -> True:True: 
Finished 5 tests.
```

That is，although a minimally strict implementation obviously yields True ：True ：$\perp$ for the argument $\perp$ ，Sloth proposes the result True ：$\perp$ as the following strict inequal－ ity illustrates．

$$
\begin{aligned}
\left(\inf _{\left.\llbracket d_{\text {definite }} \rrbracket \perp\right)\left.\right|_{\epsilon}}\right. & =\prod\left\{\llbracket \text { definite }^{\prime} \rrbracket t v \mid t v \in \llbracket[\text { Bool }] \rrbracket \uparrow, \perp \sqsubseteq t v\right\} \\
& =\llbracket \text { definite }^{\prime} \rrbracket[] \sqcap \prod\left\{\llbracket \text { definite }^{\prime} \rrbracket t v \mid t v \in \llbracket[\text { Bool }] \rrbracket \uparrow, \perp: \perp \sqsubseteq t v\right\} \\
& =\text { True }: \text { True }:[] \sqcap \prod\left\{\llbracket \text { definite }^{\prime} \rrbracket t v \mid t v \in \llbracket[\text { Bool }] \rrbracket \uparrow, \perp: \perp \sqsubseteq t v\right\} \\
& \sqsupset \text { True }: \text { True }:[] \sqcap \text { True }: \perp \\
& =\text { True }: \perp
\end{aligned}
$$

This is the price that we have to pay for identifying counter－examples as definite and for evaluating smaller parts of the tree．

### 5.3 Presenting Counter-Examples

As we have observed in Chapter 4, if we consider only sequential functions, we are looking for minimal elements with respect to the less-strict ordering. Therefore, there might be several less strict implementations for one function that are contradictory to each other.

If we check and using verboseCheck, Sloth does not present all elements of the characteristic set of and (see Example 4.3.2 for the characteristic set of and). Instead, Sloth reports the following test cases.

```
> verboseCheck and 4
1: \\perp \perp -> \perp
2: \False }\perp -> Fals
3: \True }\perp -> 
4: \False False -> False
5: \False True -> False
6: \True False -> False
7: \True True -> True
Finished 7 tests.
```

Figure 5.3.1 shows the tree that is generated by Sloth to check whether and is minimally strict. In this case the tree does not contain any witnesses that a position is non-sequential. When we consider a function whose argument type has only finitely many elements, like and, the absence of a witness implies that all remaining positions are sequential. We will later consider the case of a function whose argument type has infinitely many values.

The highlighted pairs of mapping node and result position node in Figure 5.3.1 are counter-examples. As we have observed in Example 4.3.5, there exists no sequential Boolean conjunction that satisfies both counter-examples. Instead of presenting all, potentially contradictory, counter-examples, Sloth presents one set of test cases that contains a counter-example and is completely satisfiable by a single sequential implementation.

By employing the structure of a refinement tree we can quite easily identify sets that are satisfiable by a single sequential function. For every result position node we may only consider a single argument position node. In other words, for every result position node we only keep a single successor and drop the other successors. The argument positions that remain after this process resemble the demanded positions of the proposed less strict implementation. Furthermore, the resulting tree resembles a kind of case cascade that implements the pattern matching of the less strict implementation up to a given size.

In general, we may not keep more than one argument position node per result position node as we otherwise might propose an implementation that is not sequential. For example, let us assume that we choose argument position 1 at the topmost result position node of the tree in Figure 5.3.1. In this case we are supposed to implement a function and such that $\llbracket a n d^{\prime} \rrbracket\langle$ False, $\perp\rangle \neq \perp$. By sequentiality we have $\llbracket a n d^{\prime} \rrbracket\langle\perp, v\rangle=\perp$ for all Boolean values $v$. Therefore, we have to drop all counter-examples that occur in the tree rooted at argument position 2 and do not occur in the


Figure 5.3.1: The Refinement Tree for and and two Counter-Examples.
tree rooted at argument position 1 as these are exactly the applications of the form $\llbracket a n d^{\prime} \rrbracket\langle\perp, v\rangle$ for $v \neq \perp$. We can apply the same argument to show that we may only choose a single argument position if we consider an arbitrary result position node of an arbitrary refinement tree.

Thus, in Figure 5.3.1, with respect to the outermost result position node, we have to select either the subtree that is rooted at argument position 1 or the subtree that is rooted at argument position 2 . The subtree that is rooted at argument position 1 resembles andL while the subtree that is rooted at argument position 2 resembles andR. If there are multiple subtrees that contain counter-examples, we apply the following approach. First we select the subtree that contains the smallest counter-example with respect to the number of constructors. In other words, if there are two subtrees that contain counter-examples we choose the tree whose counter-example occurs in the smaller depth. In the case of and both counter-examples occur in depth one.

If one of the argument positions in question is a demanded position, we consider the corresponding subtree. That is, in the case of and we choose argument position 1 and not position 2 because position 1 is the demanded position of $\langle\perp, \perp\rangle$ at result position $\epsilon$. Note that and first performs pattern matching on its first argument because of the left-to-right pattern matching order used by Haskell. By preferring demanded positions we keep the changes with respect to the pattern matching order minimal that have to be applied to get a less strict implementation.

If none of the argument positions in question is a demanded position, we choose the leftmost argument position. More precisely, we choose the smallest argument position with respect to a lexicographical order on positions. We think that this decision is the most natural one as it reflects the default left-to-right pattern matching order of Haskell.

### 5.4 Identifying Sequential Positions

As we have observed before, if we consider a function whose argument type has infinitely many values, the absence of a witness does not imply that a position is sequential. We might simply have to consider arguments of larger sizes to find a witness and, hence, observe that a position is non-sequential.

In this section we introduce a third highlighting, which highlights a whole test case. In contrast to previous examples, sometimes we are not able to decide whether a test case is an element of the characteristic set. Therefore, we introduce the following concept of potential test cases. Sloth uses the same highlighting as the word "potential" to indicate that a test case is a potential test case.
potential test case A potential test case might be no test case anymore if we consider inputs of larger sizes.

As an example, let us consider the following function that takes two Boolean lists and checks whether its first argument is at least as long as its second argument.

$$
\begin{aligned}
& \begin{array}{l}
\text { greaterEqual }:: ~[\text { Bool }] \rightarrow[\text { Bool }] \rightarrow \text { Bool } \\
\text { greaterEqual }[] \quad[]=\text { True } \\
\text { greaterEqual }(-:-)[] \\
\text { greaterEqual }(-: x s)(-: y s)
\end{array}=\text { True }
\end{aligned}
$$

If we check greaterEqual, Sloth yields the following counter-example that is marked as potential test case.

```
> strictCheck greaterEqual 3
2: \_\perp []_-> True
Finished 7 tests.
```

Thus, the counter-example might be no test case anymore if we consider inputs of a larger size. Therefore, the reported counter-example might be no counter-example anymore if we consider inputs of larger sizes. More precisely, there is no witness that proves that position 2 is a non-sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ if we consider arguments up to size three. Hence, position 2 might also be a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$. If it is a sequential position, then the pair $(\langle\perp,[]\rangle, \epsilon)$ is an element of the characteristic set and, thus, the presented counter-example is indeed a counter-example. If, on the other hand, position 2 is a non-sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$, then the pair $(\langle\perp,[]\rangle, \epsilon)$ is not an element of the characteristic set and, thus, the counter-example is not a coun-ter-example.

## 5 Implementation of Sloth

Nevertheless, for every non-sequential position there exists a finite witness that shows that this position is non-sequential. If a test case is not an element of the characteristic set, by checking the function for arguments up to a larger size, Sloth will eventually discover this witness and discard the test case. However, Sloth might as well never discard the test case, no matter which size we use, as the corresponding position is indeed a sequential position.

In fact, Sloth does not identify the test case for greaterEqual as a potential test case because it applies an additional criterion to identify sequential positions. The only way to identify a sequential position is to observe that it is a demanded position. For example, we have $\llbracket$ greaterEqual $\rrbracket\left\langle\perp_{1}, \perp_{2}\right\rangle=\perp_{1}$, and, hence, position 1 is a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$. So, why is position 2 a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ as well? After greaterEqual has performed pattern matching on its first argument, it performs pattern matching on its second argument. In other words, we have

$$
\llbracket \text { greaterEqual } \rrbracket\left\langle[], \perp_{2}\right\rangle=\perp_{2}
$$

and

$$
\llbracket g r e a t e r E q u a l \rrbracket\left\langle\left(\perp_{1.1}: \perp_{1.2}\right), \perp_{2}\right\rangle=\perp_{2} .
$$

That is, if we refine the first argument, in both cases position 2 becomes a demanded position. As every demanded position is also a sequential position, we have

$$
\llbracket g r e a t e r E q u a l \rrbracket\langle[], \perp\rangle=\perp
$$

and

$$
\llbracket g r e a t e r E q u a l \rrbracket\langle v, \perp\rangle=\perp
$$

for all lists $v$ with $v \sqsupseteq \perp: \perp$. As we, furthermore, have $\llbracket g r e a t e r E q u a l \rrbracket\langle\perp, \perp\rangle=\perp$, we get $\llbracket$ greaterEqual $\rrbracket \perp v \perp=\perp$ for all lists $v$. Thus, position 2 is also a sequential position in $\langle\perp, \perp\rangle$. This way, Sloth observes that the test case $(\langle\perp,[]\rangle, \epsilon)$ is definitely an element of the characteristic set of greaterEqual as position 2 is a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$.

However, even with this additional approach to identifying non-sequential positions, in some cases Sloth only reports potential test cases. For example, consider the following constant function.

```
constBool :: [Bool] \(\rightarrow\) Bool \(\rightarrow\) Bool
constBool [] \(\quad b=b\)
constBool ( \(-: b s\) ) \(b=\) constBool bs \(b\)
```

All test cases for constBool are potential elements of the characteristic set.

```
> strictCheck constBool 5
2: \\perp False -> False
3: \\perp True -> True
5: \(\perp:\perp) False -> False
7: \( }\perp:\perp) True -> True
Finished 15 tests.
```

Position 1 is a sequential position in $\langle\perp, \perp\rangle$ because it is a demanded position in $\langle\perp, \perp\rangle$, in other words, we have $\llbracket$ constBool $\rrbracket\left\langle\perp_{1}, \perp_{2}\right\rangle=\perp_{1}$. However, in contrast to greaterEqual, the function constBool does not demand its second argument after it has performed pattern matching on its first argument. We have

$$
\llbracket \text { constBool } \rrbracket\left\langle[], \perp_{2}\right\rangle=\perp_{2}
$$

but

$$
\llbracket \text { constBool } \rrbracket\left\langle\left(\perp_{1.1}: \perp_{1.2}\right), \perp_{2}\right\rangle=\perp_{1.2}
$$

More generally speaking, constBool projects to its second argument if it is applied to a list that is terminated by an empty list while it projects to the according error if the list is terminated by an error. In other words, we have

$$
\llbracket \text { constBool } \rrbracket\left\langle\left(\perp_{1.1}: \ldots:[]\right), \perp_{2}\right\rangle=\perp_{2}
$$

and

$$
\llbracket \text { constBool } \rrbracket\left\langle\left(\perp_{1.1}: \ldots: \perp_{1 . n}\right), \perp_{2}\right\rangle=\perp_{1 . n} .
$$

Therefore, Sloth is not able to observe that we have $\llbracket \operatorname{constBool\rrbracket }\langle v, \perp\rangle=\perp$ for all Boolean lists $v$ by considering finitely many test cases only. There might as exist an argument such that constBool does not demand its second argument. For example, the following definition is semantically equivalent with constBool if we consider lists up to size 2.

$$
\begin{aligned}
& \text { constBool' }::[\text { Bool }] \rightarrow \text { Bool } \rightarrow \text { Bool } \\
& \text { constBool } \left.^{\prime}(-:-)_{-}\right)=\text {True } \\
& \text { constBool }_{-}^{\prime} \quad b=b
\end{aligned}
$$

However, position 2 is not a sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$ with respect to constBool' because we have $\llbracket$ constBool $\rrbracket \backslash(\perp: \perp: \perp), \perp\rangle=$ True. In other words, as position 2 is a non-sequential position in $\langle\perp, \perp\rangle$ at result position $\epsilon$, there exists a finite witness that proves this fact, in this particular case the smallest witness is the argument $\langle(\perp: \perp: \perp), \perp\rangle$.

[^7]
## 6 Minimally Strict Polymorphic Functions

In Chapter 4 we have only considered monomorphic functions, that is, functions $f$ of type $\tau_{1} \rightarrow \tau_{2}$ where $\tau_{1}$ and $\tau_{2}$ are monomorphic types. In this section we show how we can check whether a polymorphic function is minimally strict by checking whether a specific monomorphic instance is minimally strict. More precisely, we show that we only have to check a subset of all possible inputs for the integer instance of a polymorphic function. This approach leads to a surprisingly small number of test cases. For example, when we consider functions of type $[\alpha] \rightarrow[\alpha]$ we only have to check a linear number of test cases in the number of elements in the list while there is an exponential number of test cases in the number of elements in the list for all naive monomorphic instances.

More precisely, we show that we can check whether two polymorphic functions are related by the less-strict relation by checking the functions for all inputs where we replace the polymorphic components by integers that encode the position of the component in the data structure. This approach is similar to an approach by Bernardy et al. (2010), but, in contrast, we consider non-termination and seq ${ }^{1}$. Furthermore, we show that we can use a similar approach to check whether a polymorphic function is minimally strict. In fact, this is our actual goal as we want to use the results of this chapter the improve the efficiency of testing polymorphic functions using Sloth. On the way to this efficient test of polymorphic functions we obtain several results about polymorphic functions in Haskell with/without a strict evaluation primitive. For example, we characterize a polymorphic data structure like a list by its shape and its content and prove that a polymorphic function is characterized by its behavior on shapes. This approach is similar to the container view by Bundy and Richardson (1999) and Prince et al. (2008) but, in contrast, we consider non-termination and the strict evaluation primitive seq.

In contrast to Chapter 4, in this chapter we do not strictly separate syntax and semantics because otherwise the proofs presented here become very hard to conceive. This approach is often used in the context of equational reasoning for functional programs (Gibbons 1999; Danielsson and Jansson 2004; Fernandes et al. 2007) because the semantics of a functional program is quite similar to the program itself. Note that we, in contrast to some works that use equational reasoning, still consider the influence of $\perp$, that is, the influence of run-time errors and non-terminating expressions.

In the following by $\equiv$ we denote semantic equivalence. Furthermore, we relate syntactic objects by the cpo ordering $\sqsubseteq$ and denote that their semantics are related. When we have to consider several monomorphic instances of a polymorphic func-

[^8]tion, we use a subscript type to indicate the specific instance. All proofs that are omitted in this section are found in the Appendix B.

### 6.1 Introduction

Consider the following polymorphic function called inits, which is defined in the standard Haskell library Data.List. This function takes a list and yields a list of all initial segments of this list, shortest first.

```
inits \(::[\alpha] \rightarrow[[\alpha]]\)
inits []\(=[[]]\)
inits \((x: x s)=[[]]+\operatorname{map}(x:)(\) inits \(x s)\)
```

To check this function, Sloth generates test cases, namely, elements of the argument type of inits. Therefore, we have to choose a monomorphic instance of the polymorphic function inits. Obviously, we want to choose the monomorphic instance of the polymorphic function that results in the smallest number of test cases. A candidate that might come into mind is the monomorphic instance for the unit type (), which has only a single element denoted by (). Though, unfortunately, that a monomorphic unit instance of a polymorphic function is unnecessarily strict does not imply that the corresponding polymorphic function is unnecessarily strict. For example, consider the identity function id $x=x$. The function id :: () $\rightarrow()$, that is, the monomorphic unit instance of the polymorphic identity function, is unnecessarily strict. Let us consider the unary function that emerges from applying const ${ }^{2}$ to (). The function const () :: () $\rightarrow()$ is less strict than id because we have $i d \perp \equiv \perp$ but const ()$\perp \equiv()$ and $i d() \equiv()$ as well as const ()()$\equiv()$. That is, for the argument $\perp$ the result of const () is more defined than the result of $i d$. Though, apparently, there exists no polymorphic function of type $\alpha \rightarrow \alpha$ whose unit instance behaves like const ().

As we have already mentioned in Section 3.2, to check whether a polymorphic function is as little strict as possible we instantiate all occurrences of type variables by an opaque type, called $A$, which is provided by Sloth. If we check this instance of inits for lists up to size two, Sloth reports the following counter-examples.

```
> strictCheck (inits :: [A] -> [[A]]) 2
1: \\perp -> []:\perp
3: \(a:\perp) -> []:(a:[]):\perp
Finished 5 tests.
```

The first counter-example states that inits $\perp$, yields $\perp$ while there exists a less strict implementation that yields []: $\perp$ instead. The second counter-example states that inits $(a: \perp)$, where $a$ is an arbitrary value, yields []: $\perp$ while there exists a less strict implementation that yields []: $(a:[]): \perp$ instead.

If we reconsider the implementation of inits, we observe that it checks whether its argument is the empty list although the first element of the result list is always the

[^9]empty list. If we consider the following equivalent definition of inits, this circumstance becomes even more obvious.

```
inits \(::[\alpha] \rightarrow[[\alpha]]\)
inits \(x s=\)
    case \(x s\) of
        []\(\rightarrow[]:[]\)
        \(y: y s \rightarrow[]:\) map (y:) (inits ys)
```

Thus, we can apply case deferment to derive the following less strict implementation. That is, we apply the same transformation as we have used to derive a less strict implementation of intersperse in Section 3.2. However, note that we first have to apply another transformation to inits to observe that both right-hand sides of the case expression have the same context. In Section 7.1 we will consider case deferment in more detail.

```
inits \(^{\prime}::[\alpha] \rightarrow[[\alpha]]\)
inits \({ }^{\prime} x s=\)
    [] : case \(x\) s of
        [] \(\rightarrow\) []
        \(y: y s \rightarrow \operatorname{map}(y:)\left(\right.\) inits \(\left.^{\prime} y s\right)\)
```

If we reconsider the test cases generated by Sloth, we observe that Sloth has only generated five test cases to check inits for lists up to size two. Furthermore, if we check inits for lists up to size four, Sloth generates only nine test cases. In contrast, Sloth checks exponentially many test cases in the length of the list if we test any monomorphic instance of inits. For example, if we check the Boolean instance of inits for lists up to size four, Sloth generates 21 test cases. As Sloth manages the same test with linearly many test cases, it obviously treats the data type $A$ in a special way. In this chapter we prove that the approach for $A$-instances, which is used by Sloth, is correct.

In Section 4 two functions are only related by the less-strict relation if they agree for total arguments. For simplicity, we omit this requirement here, but we can extend all results to the original definition by considering total arguments as a special case. In other words, in this chapter we employ the following definition of the less-strict relation for monomorphic functions.

Definition 6.1.1 (Less-Strict Relation, Monomorphic Functions): Let $f, g:: \tau_{1} \rightarrow \tau_{2}$ be monomorphic functions. The functions $f$ and $g$ are related by $\preceq$ if the following holds.

$$
f \preceq g: \Longleftrightarrow \forall x:: \tau_{1} . f x \sqsupseteq g x
$$

The right-hand side is equivalent to $f \sqsupseteq g$ by the semantic ordering of functions.
As this definition of the less-strict relation, as well as the original Definition 4.1.2, only consider monomorphic functions, we have to provide a generalization to polymorphic functions. Two polymorphic functions are related by the less-strict relation if and only if all their monomorphic instances are related by the less-strict relation.

Here, $f_{\tau}$ denotes the instantiation of the polymorphic function $f$ to the monomorphic type $\tau$.

Definition 6.1.2 (Less-Strict Relation, Polymorphic Functions): Let $f$ and $g$ be polymorphic functions of equal type. The functions $f$ and $g$ are related by $\preceq$ if the following holds.

$$
f \preceq g: \Longleftrightarrow \forall \tau:: * . f_{\tau} \preceq g_{\tau}
$$

By $\tau:: *$ we denote that $\tau$ is a type.
Here and in the following, we only consider non-empty types. That is, every type $\tau$ has an inhabitant besides the least element $\perp$. Finally, note that most of the results, presented in this chapter, are not restricted to considerations about the less-strict relation as we can define $f \equiv g$ by $f \preceq g \wedge g \preceq f$. For example, we can employ our results to perform the tests by Coutts et al. (2007) more efficiently. They check whether list functions that are implemented by means of their stream fusion library are equivalent to their original definition by checking whether the two implementations agree for all partial inputs. We cannot only apply the results to the area of testing but also to any form of verification that considers the influence of errors and non-terminating expressions. For example, Abel et al. (2005) have presented a transformation of Haskell programs into a monadic representation of the program, to reason about Haskell programs using the proof assistent Agda. Employing the results of the chapter at hand we can improve the efficiency of proving statements about polymorphic functions using their approach.

### 6.2 Free Theorems

The proofs in this chapter make heavy use of free theorems (Wadler 1989). Free theorems are semantic statements about functions that are derived from type information only. That is, without considering the implementation of a function we can derive semantic statements about a function. Free theorems are a mighty proof tool, which have been applied successfully in a variety of contexts (Gill et al. 1993; Johann 2002; Voigtländer 2008a,b; Oliveira et al. 2010).

Consider the following definition of the standard Haskell function reverse, which reverses a list.

```
reverse :: \([\alpha] \rightarrow[\alpha]\)
reverse \(=\) rev []
    where
        revexs [] \(=x s\)
        rev \(x s(y: y s)=r e v(y: x s) y s\)
```

For all functions $g:: \tau_{1} \rightarrow \tau_{2}$ and all lists $x s::\left[\tau_{1}\right]$ the following equation holds. By the subscript types we indicate that the two applications of reverse are, in fact, specific monomorphic instances.

$$
\text { map } g\left(\text { reverse }_{\tau_{1}} x s\right) \equiv \text { reverse }_{\tau_{2}}(\text { map } g x s)
$$

Intuitively, this equality states that it does not matter, whether we first reverse a list and apply a function to all elements of the result list afterwards or first apply a function to all elements of a list and reverse the result afterwards.

We can justify this property as follows. The function reverse takes a list and reverses the order of the elements of the list. For example, consider a list

$$
x s \equiv\left[x_{1}, \ldots, x_{n}\right]
$$

We have

$$
\text { reverse }_{\tau_{1}} x s \equiv\left[x_{n}, \ldots, x_{1}\right] .
$$

The behavior of reverse does not depend on the elements of the argument list, it only depends on the shape of the argument list. The behavior of reverse cannot depend on the elements of the list because it is polymorphic, and, therefore, it cannot inspect the elements of the argument list. Furthermore, map preserves the shape of a list. That is, reverse permutes $x s$ in the same manner as it permutes map $g x s$. More precisely, we have

$$
\text { map } g x s \equiv\left[g x_{1}, \ldots, g x_{n}\right]
$$

and

$$
\text { reverse }_{\tau_{2}}(\text { map } g x s) \equiv\left[g x_{n}, \ldots, g x_{1}\right]
$$

Therefore, we get

$$
\text { map } g\left(\text { reverse }_{\tau_{1}} x s\right) \equiv \text { reverse }_{\tau_{2}}(\text { map } g x s)
$$

Now let us consider an arbitrary function $f$ of type $[\alpha] \rightarrow[\alpha]$ instead of reverse. As $f$ is polymorphic with respect to the element type of the list, the behavior of $f$ can only depend on the shape of the list. For example, consider a list

$$
x s \equiv\left[x_{1}, \ldots, x_{n}\right]
$$

A polymorphic function like $f$ can only use the elements of the argument list as elements of the result list. But, in contrast to reverse, an arbitrary function $f$ does not have to use every element of the argument for the result, and, furthermore, it may also use elements more than once. Thus, there exists a function $i$ and a natural number $m$ such that index maps the indicies $1, \ldots, m$ of the result list to indicies $1, \ldots, n$ of the argument list and we have

$$
f_{\tau_{1}} x S \equiv\left[x_{\text {index } 1}, \ldots, x_{\text {index } m}\right]
$$

Furthermore, as map $g x s$ has the same shape as $x s$ the function $f$ chooses the same elements from the list map $g x s$ and we get

$$
f_{\tau_{2}}(\text { map } g x s) \equiv\left[g x_{\text {index } 1}, \ldots, g x_{\text {index } m}\right] .
$$

Hence, even in the more general case of an arbitrary function $f::[\alpha] \rightarrow[\alpha]$ we get

$$
\operatorname{map} g\left(f_{\tau_{1}} x s\right) \equiv f_{\tau_{2}}(\operatorname{map} g x s)
$$

The statement that for all functions $f::[\alpha] \rightarrow[\alpha]$, for all functions $g:: \tau_{1} \rightarrow \tau_{2}$ and all lists $x s::\left[\tau_{1}\right]$ we have

$$
\operatorname{map} g\left(f_{\tau_{1}} x s\right) \equiv f_{\tau_{2}}(\operatorname{map} g x s)
$$

is called the free theorem for the type $[\alpha] \rightarrow[\alpha]$. In the following we will also use the term "free theorem for $g^{\prime \prime}$, where $g$ is a function. In this case we mean the free theorem for the type of $g$ where we replace monomorphic instances of the generic function $f::[\alpha] \rightarrow[\alpha]$ by monomorphic instances of $g$. For example, the free theorem for reverse is an instance of the free theorem for the type $[\alpha] \rightarrow[\alpha]$ and states

$$
\text { map } g\left(\text { reverse }_{\tau_{1}} x s\right) \equiv \text { reverse }_{\tau_{2}}(\operatorname{map} g x s)
$$

That is, we replace monomorphic instances of the generic function $f$ in the free theorem for the type $[\alpha] \rightarrow[\alpha]$ by instances of reverse.

We can derive free theorems for all polymorphic function types. For example, consider the function

$$
\text { filter }::(\alpha \rightarrow \text { Bool }) \rightarrow[\alpha] \rightarrow[\alpha],
$$

which filters the elements of a list by a given predicate. The free theorem for filter states that for all functions $f:: \tau_{1} \rightarrow \tau_{2}$ and predicates $p:: \tau_{1} \rightarrow$ Bool, $q:: \tau_{2} \rightarrow$ Bool if we have

$$
p x \equiv q(f x)
$$

for all $x:: \tau_{1}$, then we have

$$
\text { map } f\left(\text { filter }_{\tau_{1}} p x s\right) \equiv \text { filter }_{\tau_{2}} q(\text { map } f x s)
$$

for all $x s::\left[\tau_{1}\right]$.
Throughout this chapter we employ various free theorems without stating them explicitly. The web application Automatic generation of free theorems can be used to generate these theorems from the corresponding type (Böhme 2007; Seidel and Voigtländer 2009). If the step of a proof is labeled "free theorem for $f$ ", we can generate the free theorem by inserting the type of $f$ into the web application. For example, the tool generates the following output for the type $[\alpha] \rightarrow[\alpha]$.

The theorem generated for functions of the type

$$
f: \because \forall a .[a] \rightarrow[a]
$$

in the sublanguage of Haskell with no bottoms, is:

```
\(\forall t_{1}, t_{2} \in T Y P E S, R \in R E L\left(t_{1}, t_{2}\right)\).
    \(\forall(x, y) \in \operatorname{lift}\{[]\}(R) .\left(f_{t_{1}} x, f_{t_{2}} y\right) \in \operatorname{lift}\{[]\}(R)\)
\(\operatorname{lift}\{[]\}(R)\)
\(\quad=\{([],[])\} \cup\{(x: x s, y: y s) \mid((x, y) \in R) \wedge((x s, y s) \in \operatorname{lift}\{[]\}(R))\}\)
```

Reducing all permissible relation variables to functions yields:

```
\forall\mp@subsup{t}{1}{},\mp@subsup{t}{2}{}\in\mathrm{ TYPES, g:: t1 }->\mp@subsup{t}{2}{}.
```



The most general form of a free theorem considers relations instead of functions. The first part of the output presents the free theorem in relational form. The relation $\operatorname{lift}\{[]\}(R)$ relates two lists $x$ and $y$ if and only if the lists have the same shape and their elements are point-wise related by the relation $R$. Thus, the relational free theorem states that, if the elements of two lists $x$ and $y$ are point-wise related by a relation $R$, then the $f$-images of $x$ and $y$ are also point-wise related by $R$.

In contrast to this most general form of a free theorem, the most common form of a free theorem is the instantiation of all relations with functions. That is, instead of an arbitrary relation $R$ we consider only functions. In this case the relation lift $\{[]\}(R)$ used in the relational form becomes an application of the list mapping map $f$ for a function $f$ that corresponds to the relation $R$. The functional form is more common as it is closer to the functional program, and the power of relational free theorems is only rarely necessary. Later in this chapter we will see an example where we need the power of relational free theorems to prove a statement.

As the output of the automatic generator states, the presented theorem only holds in the "sublanguage of Haskell with no bottoms", that is, in a language without runtime errors and non-termination. For example, consider the following instantiation of the free theorem for the type $[\alpha] \rightarrow[\alpha]$.

$$
\begin{aligned}
& f::[\alpha] \rightarrow[\alpha] \\
& f-=[\perp] \\
& g:: \text { Bool } \rightarrow \text { Bool } \\
& g_{-}=\text {True }
\end{aligned}
$$

If we consider the argument [ ], we have

$$
\operatorname{map} g(f[]) \equiv \operatorname{map} g[\perp] \equiv[\text { True }]
$$

while on the other hand we have

$$
f(\operatorname{map} g[]) \equiv f[] \equiv[\perp]
$$

Hence, map $g(f x s) \equiv f($ map $g x s)$ does not hold for all functions $g:: \tau_{1} \rightarrow \tau_{2}$ and all lists $x s::\left[\tau_{1}\right]$. The intuitive explanation of the free theorem fails in the presence of $\perp$ because the polymorphic function cannot only use the elements of the argument list for the result list, but it can also introduce $\perp$. Wadler (1989) has already observed that $g$ has to be strict ${ }^{3}$ if we consider run-time errors and non-termination. The automatic generation of free theorems generates the side conditions that are necessary to fix the free theorem in the presence of these features if we state that we want to consider "general recursion".

[^10]An extension of free theorems by Johann and Voigtländer (2004) even considers the influence of the strict evaluation primitive seq, which satisfies the laws seq $\perp y \equiv$ $\perp$ and seq $x y \equiv y$ if $x \not \equiv \perp$. Free theorems break in the presence of the standard Haskell function seq even if we only consider strict functions. For example, consider the following instantiation of the free theorem for the type $[\alpha] \rightarrow[\alpha]$.

$$
\begin{aligned}
& f::[\alpha] \rightarrow[\alpha] \\
& \left.f(x: y:]_{-}\right)=[\text {seq } x y] \\
& g:: \text { Bool } \rightarrow \text { Bool } \\
& g \text { True }=\text { True }
\end{aligned}
$$

If we consider the argument [False, True], we have

$$
\operatorname{map} g(f[\text { False, True }]) \equiv \text { map } g[\text { seq False True }] \equiv \text { map } g[\text { True }] \equiv[\text { True }]
$$

while on the other hand we have

$$
f(\text { map } g[\text { False, True }]) \equiv f[\perp, \text { True }] \equiv[\text { seq } \perp \text { True }] \equiv[\perp] .
$$

To fix free theorems in the presence of $s e q$ we have to assure that certain functions are total ${ }^{4}$. In the case of the free theorem above $g$ has to be total. In Section 6.3 we consider statements in a language without seq, and in Section 6.4 we consider a language with seq.

### 6.3 Less Strict Functions on Lists

In this section we consider functions of type $[\alpha] \rightarrow[\alpha]$ that do not use seq. We show that, if any monomorphic instance of two polymorphic functions is related by the less-strict relation, then all other monomorphic instances are related as well. This result implies that two polymorphic functions are already related if one of their monomorphic instances is related. Note that this does not imply that a polymorphic function is unnecessarily strict if a monomorphic instance is unnecessarily strict. As we have observed in the introduction of this chapter, there may exist a monomorphic, less strict function that does not have a polymorphic generalization.

As a first step towards the result concerning the less-strict relation, we prove for polymorphic functions $f::[\alpha] \rightarrow[\alpha]$ and $g::[\alpha] \rightarrow[\alpha]$ and an arbitrary type $\tau$ that $f_{\tau} \preceq g_{\tau}$ implies $f_{()} \preceq g_{()}$.
Lemma 6.3.1: Let $f, g::[\alpha] \rightarrow[\alpha]$. For all types $\tau$ the following holds.

$$
f_{\tau} \preceq g_{\tau} \Longrightarrow f_{()} \preceq g_{()}
$$

Intuitively, this statement is proved by employing a free theorem and relating all elements $x$ of type $\tau$ such that $x \not \equiv \perp$ with () and $\perp_{\tau}$ with $\perp_{()}$. Note that the proof employs that the type $\tau$ has an inhabitant.

[^11]Proving the converse of Lemma 6.3.1 is considerably more complex than proving Lemma 6.3.1. We start with the definition of a function shape $::[\alpha] \rightarrow[$ Int $]$ that replaces the elements of a list by their index in the list, starting with index 0 .

Definition 6.3.1 (Shape): We define the following function that replaces all elements of a list by increasing natural numbers, starting with 0 .

$$
\begin{aligned}
& \text { shape }::[\alpha] \rightarrow[\text { Int }] \\
& \text { shape }=\operatorname{zipWith}\left(\lambda n_{-} \rightarrow n\right)[0 \ldots]
\end{aligned}
$$

The expression [0..] generates an infinite list of increasing integers starting with 0 . The definition of the function zipWith can be found in Section 2.1.5.

For example, we get the results

$$
\begin{aligned}
& \text { shape }[\text { False, True, True }] \equiv[0,1,2] \text {, } \\
& \text { shape }(' z \text { ' : 'a' }: \perp) \equiv 0: 1: \perp \text {, }
\end{aligned}
$$

and

$$
\text { shape }[\perp, 8] \equiv[0,1] \text {. }
$$

Note that $0: 1: \perp$ is different from $[0,1, \perp]$ as the former one denotes a list that contains the elements 0 and 1 and is terminated by $\perp$ while the latter one denotes a list that contains the elements 0,1 , and $\perp$ and is terminated by the empty list. To be more precise we could as well write $0: 1: \perp_{[\text {Int }}$ and $\left[0,1, \perp_{\text {Int }}\right]$.

The function shape is similar to the function template defined by Voigtländer (2009). While shape yields only the shape of a list, template additionally yields a mapping from indices to the original elements of the list. By using this mapping it is possible to reconstruct the original list. Here, we define the mapping by means of the list indexing function (!!) :: $[\alpha] \rightarrow$ Int $\rightarrow \alpha$. The function (!!) takes a list and an index in the list (where the smallest index is 0 ) and yields the element at the corresponding position. If the index does not exist, the function yields $\perp$ instead. The definition of (!!) is presented in Section 2.1.4. More importantly than the fact that we use list indexing, shape can handle partially defined lists, while template cannot. For example, we have template $(3: 10: \perp) \equiv \perp$. Note that this difference is essential as we want to use shape to prove statements about the less-strict relation.

To prove that we can characterize a list by its shape and (!!) we prove the following lemma. The characterization of a list by means of shape and (!!) is similar to the container approach by Prince et al. (2008). As Prince et al. (2008) do not consider $\perp$ or seq, a list is represented by its length (that is, the shape, appearing as the parameter of a dependent type) and a function from index to element.

Lemma 6.3.2: For all $x:: \tau$ and $x s::[\tau]$ the following holds.

$$
\operatorname{map}((x: x s)!!)[1 . .] \equiv \operatorname{map}(x s!!)[0 \ldots]
$$

Although this statement looks quite obvious, we have to use a relational free theorem to prove it. We cannot use a standard free theorem as we have to somehow incorporate that $(x: x s)!!$ on the left-hand side is never applied to the index 0 .

Now we are ready to show that a list is reconstructible from its shape by employing list indexing. In the following proof we employ the free theorem for zipWith, which can be generated by entering "zipWith" into the free theorem generator. The theorem states that for all functions $f, g, h, p$, and $q$ of appropriate type that satisfy

$$
h(p x y) \equiv q(f x)(g y)
$$

for all values $x:: \tau_{1}$ and $y:: \tau_{2}$, we have

$$
\text { map } h(z i p \text { With } p z v) \equiv \text { zipWith } q(\text { map } f z)(\operatorname{map} g v)
$$

for all lists $z::\left[\tau_{1}\right]$ and $v::\left[\tau_{2}\right]$. We give a detailed proof of the following lemma as we reconsider it in the presence of seq.

Lemma 6.3.3: For all lists $x s::[\tau]$ the following holds.

$$
\operatorname{map}(x s!!)(\text { shape } x s) \equiv x s
$$

Proof: Proof by structural induction over $x$ s.
Base Cases: If $x s \equiv \perp$ or $x s \equiv[]$, we reason as follows.

```
map (xs!!) (shape (\perp / []))
    \equiv { definition of shape }
map (xs!!) (zipWith (\lambdan_ ->n)[0..] (\perp/[]))
\equiv { definition of zipWith }
map (xs!!) ( }\perp/[]
{ {definition of map }
\perp / []
```

Inductive Step: If $x s \equiv y: y s$, we reason as follows.

```
map ((y:ys)!!) (shape (y:ys))
    \equiv{definition of shape }
map ((y:ys)!!)(zipWith (\lambdan_ }->n)[0..] (y:ys)
    \equiv{definition of [0..]}
map ((y:ys)!!) (zipWith (\lambdan_ }->n)(0:[1..])(y:ys)
    \equiv { definition of zipWith }
map ((y:ys)!!) (0:zipWith (\lambdan_ 
\equiv{definition of map }
((y:ys)!! 0) :map ((y:ys)!!) (zipWith (\lambdan_}->n)[1..] ys
    \equiv{definition of (!!)}
y:map ((y:ys)!!) (zipWith (\lambdan_ }->n)[1..] ys
    \equiv{free theorem for zipWith,((y:ys)!!) strict }
y:zipWith (\lambdan_ 
    \equiv {Lemma 6.3.2 }
y:zipWith (\lambdan_ <n) (map (ys!!) [0..]) ys)
\equiv{ free theorem for zipWith, (ys!!) strict }
```

$$
\begin{aligned}
& y: \text { map }(y s!!)(z i p W i t h ~(\lambda n-\rightarrow n)[0 . .] y s) \\
& \equiv\{\text { induction hypothesis }\} \\
& y: y s
\end{aligned}
$$

To show the use of a free theorem in detail, we exemplarily state the instantiation of the free theorem for zipWith to prove the step labeled with $(*)$.

$$
\begin{array}{lll}
h=((y: y s)!!) & q=\lambda n-\rightarrow n & z=[1 \ldots] \\
p=\lambda n-\rightarrow n & f=((y: y s)!!) & v=y s \\
g=i d & &
\end{array}
$$

It is easy to show that we have $h(p x y) \equiv q(f x)(g y)$ for all appropriate $x$ and $y$, which proves the step labeled with $(*)$. We return to the steps labeled with $(*)$ and $(* *)$ when we consider the presence of seq.

The just proved characterization of lists directly allows for a characterization of polymorphic list functions via their behavior on shapes. The Int instance of the polymorphic function $f$ may be compared with a corresponding container morphism of $f$ (Prince et al. 2008).

The following proof employs the free theorem for functions of type $[\alpha] \rightarrow[\alpha]$, which states that we have $f_{\tau_{2}}(\operatorname{map} g x s) \equiv \operatorname{map} g\left(f_{\tau_{1}} x s\right)$ for all strict functions $g:: \tau_{1} \rightarrow \tau_{2}$ and all $x s::\left[\tau_{1}\right]$.

Lemma 6.3.4: For all functions $f::[\alpha] \rightarrow[\alpha]$ and all lists $x s::[\tau]$ the following holds.

$$
f_{\tau} x s \equiv \operatorname{map}(x s!!)\left(f_{\text {Int }}(\text { shape } x s)\right)
$$

Proof: Let $x s::[\tau]$. We reason as follows.

```
f\tau}x
    \equiv {Lemma 6.3.3}
f
    \equiv {free theorem for f, (xs!!) strict }
map (xs!!) (f flnt (shape xs))
```

We return to the step labeled with $(*)$ when we consider the presence of seq.
Lemma 6.3.4 enables us to prove a kind of standard lemma about functions of type $[\alpha] \rightarrow[\alpha]$. It is a formalization of the intuitive explanation of the validity of free theorems.

A polymorphic function of type $[\alpha] \rightarrow[\alpha]$ can only use the elements of its argument list and $\perp$ for the elements of the result list. Therefore, for every element in the result list there exists a position in the argument list, where it is taken from, or it is $\perp$. Furthermore, because the function is polymorphic, it can distinguish argument lists only by their shape. Thus, if we apply a polymorphic function to two lists with equal shape it chooses elements from the same positions of the argument list for the result list.

The following proof employs an instance of the free theorem for functions of type $[\alpha] \rightarrow \alpha$. More precisely, for a number $n::$ Int we consider the function (!!n) :: $[\alpha] \rightarrow$ $\alpha$. The corresponding free theorem states that for all strict functions $g:: \tau_{1} \rightarrow \tau_{2}$ and all $x s::\left[\tau_{1}\right]$ we have (map $\left.g x s\right)!!n \equiv g(x s!!n)$.

Lemma 6.3.5: Let $f::[\alpha] \rightarrow[\alpha]$ and $x s::\left[\tau_{1}\right]$ and $y s::\left[\tau_{2}\right]$ such that shape $x s \equiv$ shape $y s$. Then, for all $n::$ Int there exists $k::$ Int with

$$
\left(f_{\tau_{1}} x s\right)!!n \equiv x s!!k \text { and }\left(f_{\tau_{2}} y s\right)!!n \equiv y s!!k .
$$

Proof: Let $x s::\left[\tau_{1}\right]$ and $n::$ Int. We define $k$ as $\left(f_{\text {Int }}(\right.$ shape $\left.x s)\right)!!n$ and reason as follows.

```
\(\left(f_{\tau_{1}} x s\right)!!n\)
    \(\equiv \quad\{\) Lemma 6.3.4 \(\}\)
\(\left(\right.\) map \((x s!!)\left(f_{\text {Int }}(\right.\) shape \(\left.\left.x s)\right)\right)!!n\)
\(\equiv \quad\{\) free theorem for (!!n), (xs!!) strict \}
\(x s!!\left(\left(f_{\text {Int }}(\right.\right.\) shape \(\left.\left.x s)\right)!!n\right)\)
\(\equiv\{\) definition of \(k\}\)
\(x s!!k\)
```

Let $y s::\left[\tau_{2}\right]$ such that shape $x s \equiv$ shape $y s$. We reason as follows.

```
\(\left(f_{\tau_{2}} y s\right)!!n\)
    \(\equiv\) \{ like above \}
ys !! (( \(f_{\text {Int }}(\) shape ys \(\left.\left.)\right)!!n\right)\)
    \(\equiv \quad\{\) shape \(x s \equiv\) shape \(y s\), definition of \(k\}\)
ys!! \(k\)
```

Note that, if $n$ is out of bounds, then $\left(f_{\text {Int }}(\right.$ shape $\left.x s)\right)!!n \equiv \perp$, that is, $k \equiv \perp$.
Back on the road, the following lemma is the essential lemma to prove that $f_{()} \preceq$ $g_{()}$implies $f_{\tau} \preceq g_{\tau}$ for all types $\tau$, or, more precisely, to prove its contraposition, that is, $f_{\tau} \npreceq g_{\tau}$ implies $f_{()} \npreceq g_{()}$. This proof relies on mapping a list $x s::[\tau]$ such that $f_{\tau} x s \nsupseteq g_{\tau} x s$ to a list $y s::[()]$ such that $f_{()} y s \nexists g_{()} y s$. But, constructing a suitable map, we encounter the following problem. When we consider an arbitrary type $\tau$, the type $[\tau]$ may contain values $x s$ and $y s$ such that $x s!!k \not \equiv \perp, y s!!k \not \equiv \perp$, and $x s!!k \sqsupset y s!!k$. This case does not occur for values of type $[()]$ as the elements of a list of type [()] are either $\perp$ or (). By the following lemma we elude this problem. It shows that for every problematic list there exists a non-problematic list that can be considered instead.

Lemma 6.3.6: Let $f, g::[\alpha] \rightarrow[\alpha]$. For every $x s::[\tau]$ and $n::$ Int such that

$$
\left(f_{\tau} x s\right)!!n \not \equiv \perp,\left(g_{\tau} x s\right)!!n \not \equiv \perp, \text { and }\left(f_{\tau} x s\right)!!n \not \equiv\left(g_{\tau} x s\right)!!n
$$

there exists ys :: $[\tau]$ such that

$$
\left(f_{\tau} y s\right)!!n \equiv \perp \text { and }\left(g_{\tau} y s\right)!!n \not \equiv \perp
$$

Now we are ready to prove that $f_{\tau} \npreceq g_{\tau}$ implies $f_{()} \npreceq g_{()}$for all types $\tau$. We employ that we have $x \mathrm{~s} \nexists y s$ if and only if shape $x \mathrm{~s} \not \equiv$ shape ys or $x s!!n \nexists y s!!n$ for some $n::$ Int. Furthermore, if there exists a list $x s::[\tau]$ such that $\left(f_{\tau} x s\right)!!n \not \equiv \perp$, $g_{\tau} x s!!n \not \equiv \perp$, and $\left(f_{\tau} x s\right)!!n \not \equiv\left(g_{\tau} x s\right)!!n$ by the previous lemma we can instead consider a list $y s::[\tau]$ such that $\left(f_{\tau} y s\right)!!n \equiv \perp$ and $\left(g_{\tau} y s\right)!!n \not \equiv \perp$.

Lemma 6.3.7: Let $f, g::[\alpha] \rightarrow[\alpha]$. If there exists a type $\tau$ such that $f_{\tau} \npreceq g_{\tau}$, then $f_{()} \npreceq g_{()}$.
Let us summarize the statements so far. If for any type the monomorphic instances of two polymorphic funcitons are related by the less-strict relation, then for all other types the monomorphic instances are related by the less-strict relation as well.

Theorem 6.3.1: Let $f, g::[\alpha] \rightarrow[\alpha]$. For all types $\tau_{1}$ and $\tau_{2}$ the following holds.

$$
f_{\tau_{1}} \preceq g_{\tau_{1}} \Longleftrightarrow f_{\tau_{2}} \preceq g_{\tau_{2}}
$$

Proof: By Lemma 6.3.1 and Lemma 6.3.7 we have $f_{\tau} \preceq g_{\tau} \Longleftrightarrow f_{()} \preceq g_{()}$for all types $\tau$. We prove the statement by instantiations of $\tau$ to $\tau_{1}$ and $\tau_{2}$.

Employing Theorem 6.3.1 we can check whether two polymorphic functions are related by the less-strict relation by checking whether the unit instances are related by the less-strict relation. But, the characterization of lists by means of shape allows for a statement that is even more powerful than Theorem 6.3.1 with respect to exhaustive testing. We prove that it suffices to check two polymorphic functions for all possible images of shape whether they are related by the less-strict relation. Note that, if we consider lists of length $n$, then there are only linearly many shapes while there are exponentially many lists of type $[\tau]$ for any non-empty type $\tau$.

Theorem 6.3.2: For $f, g::[\alpha] \rightarrow[\alpha]$ the following statement holds.

$$
f_{\text {Int }} \circ \text { shape } \preceq g_{\text {Int }} \circ \text { shape } \Longleftrightarrow f \preceq g
$$

Proof: $\Longrightarrow$ : Let $x s::[\tau]$. We reason as follows.

```
\(f_{\tau} x s\)
    \(\equiv \quad\{\) Lemma 6.3.4 \(\}\)
    map \((x s!!)\left(f_{\text {Int }}(\right.\) shape \(\left.x s)\right)\)
    \(\sqsupseteq \quad\left\{\right.\) monotonicity, \(f_{\text {Int }} \circ\) shape \(\preceq g_{\text {Int }} \circ\) shape \(\}\)
    map \((x s!!)\left(g_{\text {Int }}(\right.\) shape \(\left.x s)\right)\)
    \(\equiv \quad\{\) Lemma 6.3.4 \(\}\)
    \(g_{\tau} x s\)
```

$\Longleftarrow$ : We prove the contraposition of this implication. If there exists $x s::[\tau]$ such that $f_{\text {Int }}$ (shape $\left.x s\right) \nsupseteq g_{\text {Int }}$ (shape $x s$ ), then there exists a list is :: [Int], namely, is = shape $x s$, such that $f_{\text {Int }}$ is $\nexists g_{\text {Int }}$ is. Thus, we have $f \npreceq g$ by the definition of the less-strict relation for polymorphic functions (Definition 6.1.2).

### 6.4 Less Strict Functions in the Presence of seq

In this section we consider a language that provides a strict evaluation primitive seq. Although Theorem 6.3.1, proved in the absence of seq, seems quite natural, the statement breaks when seq is available. In the presence of seq the property $f_{()} \preceq g_{()}$ does not imply $f_{\tau} \preceq g_{\tau}$ for all types $\tau$. For example, consider the following functions of type $[\alpha] \rightarrow[\alpha]$.

$$
\begin{aligned}
& f::[\alpha] \rightarrow[\alpha] \\
& f\left(x: y: \__{-}\right)=[\operatorname{seq} x y] \\
& g::[\alpha] \rightarrow[\alpha] \\
& g\left(x: y:{ }_{-}\right)=[\operatorname{seg} y x]
\end{aligned}
$$

If one of the first two elements of the argument list is $\perp$, then both functions yield $[\perp]$. If both elements are (), then both functions yield [()]. That is, we have $f_{()} \preceq g_{()}$, or, more precisely, we even have $f_{()} \equiv g_{()}$. However, there are obviously types $\tau$ such that $f_{\tau}$ and $g_{\tau}$ are incomparable. For example, we have $f_{\text {Bool }}$ [False, True] $\equiv$ [False] but $g_{\text {Bool }}[$ False, True $] \equiv[$ True $]$. This example shows that Theorem 6.3.1 fails in the presence of seq.

So, how about Theorem 6.3.2, does it break as well? It holds for the previous example, but it breaks for the following functions.

$$
\begin{aligned}
& f^{\prime}::[\alpha] \rightarrow[\alpha] \\
& f^{\prime}\left(x: y: z:{ }_{-}\right)=[\operatorname{seq} x z] \\
& g^{\prime}::[\alpha] \rightarrow[\alpha] \\
& g^{\prime}\left(x: y: z:{ }_{-}\right)=\left[\begin{array}{ll}
\operatorname{seq} y & z
\end{array}\right]
\end{aligned}
$$

The functions $f^{\prime}$ and $g^{\prime}$ act identically on shapes but are incomparable with respect to the less-strict relation. Hence, Theorem 6.3.2 is also not valid if we consider seq.

In contrast to Theorem 6.3.1 the assertion of Theorem 6.3.2 directly depends on the notion of shape. Thus, maybe we can prove a similar statement when we alter shape. Let us track back where the current proof is corrupted by seq. We start with the proof of Lemma 6.3.4.

Lemma 6.3.3 is still valid because none of the considered functions uses seq. In this case we are quasi considering a language without seq. In contrast, Lemma 6.3.4 considers an arbitrary function $f$ that might use seq. According to Johann and Voigtländer (2004), the step labeled with ( $\dagger$ ) in the proof of Lemma 6.3.4 might not hold in the presence of seq as ( $x s!!$ ) is not total ${ }^{5}$. In fact, the lemma fails completely. For example, consider the function $f$ again. We have

$$
f[\perp, \text { True }] \equiv[\perp],
$$

and, on the other hand, the following holds.

[^12]```
map \(([\perp\), True \(]!!)(f(\) shape \([\perp\), True \(]))\)
    \(\equiv\{\) definition of shape \(\}\)
map \(([\perp\), True \(]!!)(f[0,1])\)
    \(\equiv\{\) definition of \(f\}\)
map ([ \(\perp\),True \(]!!\) ) [1]
    \(\equiv\{\) definition of map \(\}\)
[ \([\perp\), True] !! 1]
    \(\equiv\{\) definition of \((!!)\}\)
[True]
```

That is, we have $f x$ s $\not \equiv$ map $(x s!!)(f($ shape $x s))$.
The function $f$ is able to distinguish the list $[\perp$, True $]$ from the list $[0,1]$ by employing seq, although these lists have the same shape. Hence, in the presence of seq a polymorphic function cannot only distinguish two lists by their shape in the sense of the function shape. It can also distinguish them by the occurrences of $\perp$ as list element. Thus, to take seq into account we have to consider a different notion of shape, implemented by the function shape ${ }_{\text {seq }}$.

$$
\begin{aligned}
& \text { shape }_{\text {seq }}::[\alpha] \rightarrow[\text { Int }] \\
& \text { shape }_{\text {seq }}=\operatorname{zipWith}(\lambda n x \rightarrow \operatorname{seq} x n)[0 . .]
\end{aligned}
$$

The function shape ${ }_{\text {seq }}$ replaces all elements $x$ of a list with $x \not \equiv \perp$ by their position in the list. However, in contrast to shape, shape seq keeps all elements $x$ with $x \equiv \perp$. For example, we have

$$
\text { shape }_{\text {seq }}[\text { False }, \perp, \text { True }] \equiv[0, \perp, 2]
$$

while

$$
\text { shape }[\text { False }, \perp, \text { True }] \equiv[0,1,2] \text {. }
$$

In the remaining part of this section we prove that a theorem like Theorem 6.3.2 holds in the presence of seq if we replace shape by shape ${ }_{s e q}$. The proofs in particular highlight the role of free theorems.

If we replace shape by shape ${ }_{\text {seq }}$ in Lemma 6.3.4 we can provide a characterization of polymorphic functions in the presence of seq. To prove the characterization, first we have to ensure that Lemma 6.3.3 also holds for shape ${ }_{\text {seq }}$. This is not obvious, because shape ${ }_{\text {seq }}$ uses seq, and consequently its free theorem is more restrictive. We employ an extension of free theorems by Seidel and Voigtländer (2009) to manage the proof. According to Johann and Voigtländer (2004), in the presence of seq, we have to guarantee that certain functions are total. However, the extension by Seidel and Voigtländer (2009) allows for less restrictive free theorems. Given a function definition, it assigns a refined type to the function and uses this refined type to derive less restrictive free theorems.

To generate a free theorem for a specific implementation of a polymorphic function we have to define the function in a simple functional language with a fixpoint operator and a seq primitive. Figure 6.4.1 shows the free theorem that is generated

```
\(\forall f:: \tau_{1} \rightarrow \tau_{2}, f\) strict and total.
    \(\forall g:: \tau_{3} \rightarrow \tau_{4}, g\) strict.
        \(\forall x s::\left[\tau_{3}\right] . t x s \not \equiv \perp \Longleftrightarrow t(\) map \(g x s) \not \equiv \perp\)
            \(\wedge \forall y s::\left[\tau_{1}\right]\). map \(g(t x s y) \equiv t(\) map \(g x s)(\) mapf \(y s)\)
```

Figure 6.4.1: The Specialized Free Theorem where $t \equiv \operatorname{zipWith}(\lambda x n \rightarrow \operatorname{seq} x n)$.
for zipWith $(\lambda n x \rightarrow \operatorname{seq} x n)::[\alpha] \rightarrow[\beta] \rightarrow[\alpha] .{ }^{6}$ By inspecting the implementation the generator observes that we only use seq on the second argument of the function passed to zipWith. Therefore, in contrast to the general free theorem, the specialized version does not require $g$ to be total. This enables us to prove the following adjusted version of Lemma 6.3.3.

Lemma 6.4.1: For all lists $x s::[\tau]$ the following holds.

$$
\operatorname{map}(x s!!)\left(\text { shape }_{\text {seq }} x s\right) \equiv x s
$$

Proof: We reason the same way as in Lemma 6.3 .3 but replace all occurrences of shape by shape ${ }_{\text {seq }}$ and all occurrences of $(\lambda n-\rightarrow n)$ by $(\lambda n x \rightarrow$ seq $x n)$. Instead of the general free theorem for zipWith we employ the specialized free theorem for zipWith $(\lambda n x \rightarrow$ seq $x n$ ) shown in Figure 6.4.1. We replace the steps labeled with $(*)$ and $(* *)$ in the proof of Lemma 6.3.3 by the following reasoning. We prove the step labeled with $(*)$ by instantiating the specialized free theorem as follows.

$$
\begin{array}{ll}
g=((y: y s)!!) & x=[1 . .] \\
f=i d & y=y s
\end{array}
$$

We prove the step labeled with $(* *)$ by instantiating the specialized free theorem as follows.

$$
\begin{array}{ll}
g=(y s!!) & x=[0 . .] \\
f=i d & y=y s
\end{array}
$$

Note that id is strict and total and that $((y: y s)!!)$ and (ys!!) are strict.
To complete the proof of an adjusted version of Lemma 6.3 .4 with shape ${ }_{\text {seq }}$, we still have to recover the step labeled with $(\dagger)$ in the original proof. That is, we want to prove that

$$
f\left(\text { map }(x s!!)\left(\operatorname{shape}_{\text {seq }} x s\right)\right) \equiv \operatorname{map}(x s!!)\left(f\left(\text { shape }_{\text {seq }} x s\right)\right) .
$$

If we consider Johann and Voigtländer (2004), we have to show that ( $x s!!$ ) is total, which is not the case. First of all the position we are projecting to might be out of bounds, for example, $[1,2]!!3$. Furthermore, the list $x$ s may contain $\perp$, which

[^13]also breaks totality if we project to the corresponding position, for example, $[1, \perp]$ !! 1. We cannot employ the extension by Seidel and Voigtländer (2009) because we do not consider a concrete function like zipWith before but an arbitrary function of type $[\alpha] \rightarrow[\alpha]$. Thus, we cannot weaken the requirements of a free theorem by providing a concrete implementation. Nevertheless, there is rescue. For a specific list the requirements proposed by Johann and Voigtländer (2004) are unnecessarily strong. For example, for the free theorem
$$
\operatorname{map} g(f x s) \equiv f(\operatorname{map} g x s)
$$
they demand that $g$ has to be total. However, it suffices that $g$ is total with respect to the elements of the list $x$ s. The following theorem proves this observation by employing the relational version of the free theorem for the type $[\alpha] \rightarrow[\alpha]$.

Theorem 6.4.1: Let $f::[\alpha] \rightarrow[\alpha]$. For all strict functions $g:: \tau_{1} \rightarrow \tau_{2}$ and for all $x s::\left[\tau_{1}\right]$ with

$$
g(x s!!n) \equiv \perp \Longleftrightarrow x s!!n \equiv \perp
$$

for all $n::$ Int, we have

$$
\operatorname{map} g\left(f_{\tau_{1}} x s\right) \equiv f_{\tau_{2}}(\operatorname{map} g x s)
$$

Proof: Let $x s::\left[\tau_{1}\right]$. We define a relation that relates all elements of the list $x s$ with the image of $g$ of this element and use the relational free theorem for $f$ to prove the statement. As $R_{x s}$ has to be a continuous relation, we close the relation by relating the suprema of all possible chains.

$$
R_{x s}:=\left\{\left(\bigsqcup_{i \in I}(x s!!i), \bigsqcup_{i \in I} g(x s!!i)\right) \mid\langle x s!!i\rangle_{i \in I} \text { chain, } I \subseteq \text { Int }\right\}
$$

In particular we have $(x s!!i, g(x s!!i)) \in R_{x s}$ for all $i::$ Int as the set $\{i\}$ is a chain.
First we show that $R_{x s}$ is a strict, bottom-reflecting and continuous relation. For details about these concepts consider Definition B.1.2 in the Appendix. By definition of $R_{x s}$ we have $(x s!!\perp, g(x s!!\perp)) \in R_{x s}$. Furthermore, we have $x s!!\perp \equiv \perp$ as well as $g(x s!!\perp) \equiv g \perp \equiv \perp$, that is, $R_{x s}$ is strict.
Next we show that $R_{x s}$ is bottom-reflecting. Let $(x, y) \in R_{x s}$. By definition of $R_{x s}$ we have $x \equiv \bigsqcup_{i \in I}(x s!!i)$ and $y \equiv \bigsqcup_{i \in I} g(x s!!i)$ and reason as follows.

$$
\begin{aligned}
\bigsqcup_{i \in I}(x s!!i) \equiv \perp & \Longleftrightarrow \forall i \in I . x s!!i \equiv \perp \\
& \Longleftrightarrow \forall i \in I . g(x s!!i) \equiv \perp \\
& \Longleftrightarrow \bigsqcup_{i \in I} g(x s!!i) \equiv \perp
\end{aligned}
$$

Finally, we have to show that $R_{x s}$ is continuous. Let $\left\langle x_{i}\right\rangle_{i \in I}$ and $\left\langle y_{i}\right\rangle_{i \in I}$ be chains whose elements are pair-wise related by $R_{x s}$. Then, for every index $i \in I$ there is an index-set $I_{i}$ such that $x_{i}=\bigsqcup_{j \in I_{i}}\left(x_{s}!!j\right)$ and $y_{i}=\bigsqcup_{j \in I_{i}}\left(g\left(x_{s}!!j\right)\right)$. In the following,
we use the set $K:=\left\{j \mid i \in I, j \in I_{i}\right\}$. We have

$$
\left(\bigsqcup_{k \in K}(x s!!k), \bigsqcup_{k \in K} g(x s!!k)\right)=\left(\bigsqcup_{i \in I} x_{i}, \bigsqcup_{i \in I} y_{i}\right),
$$

and, by definition of $R_{x s}$, the left-hand side is an element of $R_{x s}$. Thus $R_{x s}$ is continuous.

We have $(x s$, map $g x s) \in \operatorname{lift}\{[]\}\left(R_{x s}\right)$ where $\operatorname{lift}\{[]\}(R)$ is the relational structural lifting defined as follows.

$$
\operatorname{lift}\{[]\}(R)=\{(\perp, \perp),([],[])\} \cup\{(x: x s, y: y s) \mid(x, y) \in R,(x s, y s) \in \operatorname{lift}\{[]\})(R)\}
$$

Intuitively $(x s, y s) \in \operatorname{lift}\{[]\}\left(R_{x s}\right)$ means that $y s$ is the point-wise $g$-image of $x s$ or, more precisely, $y s \equiv$ map $g x s$. Because the relational free theorem states that $(x s$, map $g x s) \in \operatorname{lift}\{[]\}\left(R_{x s}\right)$ implies $(f x s, f($ map $g x s)) \in \operatorname{lift}\{[]\}\left(R_{x s}\right)$, we obtain that $f$ (map $g x s)$ is the point-wise $g$-image of $f x$, that is, map $g(f x s) \equiv$ $f$ (map $g x s$ ).

For proving statements about functions in the presence of seq the previous theorem is quite useful as we are not restricted to total functions anymore. While the previous theorem does only consider functions of type $[\alpha] \rightarrow[\alpha]$, by employing the generalization sketched in Section 6.6, we can generalize this theorem to arbitrary first-order type constructors.

Additionally, this kind of generalization of a functional free theorem is also valuable in the absence of seq. For example, the standard functional free theorem for a function $f::(\alpha \rightarrow \beta) \rightarrow[\alpha] \rightarrow[\beta]$ states that

$$
h(p x) \equiv q(g x)
$$

for all $x$ of appropriate type implies

$$
\operatorname{map} h(f p x s) \equiv f q(\operatorname{map} g x s)
$$

for all $x s$ of appropriate type. By employing the relation $R_{x s}$ defined in the previous proof we can derive a free theorem that states for a list $x s$ that

$$
h(p(x s!!n)) \equiv q(g(x s!!n))
$$

for all $n::$ Int already implies

$$
\operatorname{map} h(f p x s) \equiv f q(\operatorname{map} g x s)
$$

That is, we only have to prove the relation between $h, p, q$, and $g$ for the elements of the list $x s$ and not for all $x$ of appropriate type. Actually, we have used exactly this statement to prove Lemma 6.3.2.

We want to employ Theorem 6.4.1 to prove

$$
\operatorname{map}(x s!!)\left(f\left(\text { shape }_{\text {seq }} x s\right)\right) \equiv f\left(\operatorname{map}(x s!!)\left(\text { shape }_{\text {seq }} x s\right)\right)
$$

to fix step $(\dagger)$ in the proof of Lemma 6.3.4. Thus, we have to show that we have $x s!!\left(\left(\right.\right.$ shape $\left.\left._{\text {seq }} x s\right)!!n\right) \equiv \perp$ if and only if $\left(\right.$ shape $\left._{\text {seq }} x s\right)!!n \equiv \perp$. Therefore, we show a connection between (!!) and zipWith, namely, that

$$
(z i p W i t h f x s y s)!!n \equiv f(x s!!n)(y s!!n) .
$$

This statement is quite similar to the free theorem for (!!), which states that

$$
(\operatorname{map} f x s)!!n \equiv f(x s!!n) .
$$

Indeed, these statements are closely related. In the context of free theorems, mostly, only two-ary logical relations are considered. The connection between zipWith and (!!) is the functional instantiation of the relational free theorem for (!!) if we consider three-ary relations. Alternatively, we can prove the following lemma by means of structural induction.

Lemma 6.4.2: For all strict ${ }^{7}$ functions $f:: \tau_{1} \rightarrow \tau_{2} \rightarrow \tau_{3}$ and all lists $x s::\left[\tau_{1}\right], y s::\left[\tau_{2}\right]$ the following holds.

$$
(z i p W i t h f x s y s)!!n \equiv f(x s!!n)(y s!!n)
$$

By employing the previous lemma we prove a characteristic property of shape seq , which is not valid for shape. The shape of a list with respect to shape ${ }_{\text {seq }}$ is $\perp$ at a certain position if and only if the "original list" is $\perp$ at that position.

Lemma 6.4.3: For all lists $x s::[\tau]$ and all $n::$ Int we have

$$
\left(\text { shape }_{\text {seq }} x s\right)!!n \equiv \perp \Longleftrightarrow x s!!n \equiv \perp
$$

Proof: We start with the following calculation.

```
(shape seq }xs)!!
    \equiv{definition of shape seq }
(zipWith (\lambdanx->seq x n) [0..] xs)!!n
    \equiv {Lemma 6.4.2, seq strict }
seq (xs !!n) ([0..]!!n)
```

If we have $n \equiv \perp$ or $n<0$, then seq $(x s!!n)([0 .]!!n.) \equiv \operatorname{seq} \perp \perp \equiv \perp$ as well as $x s!!n \equiv \perp$. If we have $n>=0$, then $\operatorname{seq}(x s!!n)([0 \ldots]!!n) \equiv \operatorname{seq}(x s!!n) n$. That is, $\left(\right.$ shape $\left._{\text {seq }} x s\right)!!n \equiv \perp$ if and only if $x s!!n \equiv \perp$.

Employing the characterization of shape $_{\text {seq }}$ we prove the analogue to Lemma 6.3.4 in the presence of seq.

Lemma 6.4.4: For all functions $f::[\alpha] \rightarrow[\alpha]$ and all lists $x s::[\tau]$ we have the following.

$$
f_{\tau} x s \equiv \operatorname{map}(x s!!)\left(f_{\text {Int }}\left(\text { shape }_{\text {seq }} x s\right)\right)
$$

[^14]Proof: We reason as follows.

```
\(f_{\tau} x s\)
    \(\equiv \quad\{\) Lemma 6.4.1 \(\}\)
\(f_{\tau}\left(\operatorname{map}(x s!!)\left(\right.\right.\) shape \(\left.\left._{\text {seq }} x s\right)\right)\)
    \(\equiv\) \{ Theorem 6.4.1, Lemma 6.4.3 \}
map \((x s!!)\left(f_{\text {Int }}\left(\right.\right.\) shape \(\left.\left._{\text {seq }} x s\right)\right)\)
```

This proof is very similar to the proof of Lemma 6.3.4, but instead of a general free theorem we use the specialized version proved in Theorem 6.4.1.

Now we have the means at hand to prove the analogue of Theorem 6.3.2 in the presence of seq. The proof of this statement is analogous to the proof of Theorem 6.3.2, but instead of Lemma 6.3.4 we employ Lemma 6.4.4.

Theorem 6.4.2: For all $f, g::[\alpha] \rightarrow[\alpha]$ the following holds.

$$
f_{\text {Int }} \circ \text { shape }_{\text {seq }} \preceq g_{\text {Int }} \circ \text { shape }_{\text {seq }} \Longleftrightarrow f \preceq g
$$

The previous lemma states that we can check whether a function $f$ is less strict than a function $g$ in the presence of seq by checking them for all images of shape ${ }_{s e q}$. In fact, we have observed that, in this setting, we probably can reduce the number of test cases significantly. More precisely, in the following we illustrate that it is possible to consider linearly many test cases in the length of a list instead of exponentially many test cases in the length of the list to distinguish two functions of type $[\alpha] \rightarrow[\alpha]$ in the presence of seq.

Let us consider a function of type $\alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$ in a language without seq. By employing free theorems we can show that every function $f:: \alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$ is semantically equivalent to one of the following functions.

$$
\begin{aligned}
& f_{1}:: \alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha \\
& f_{1} x y z=\perp \\
& f_{2}:: \alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha \\
& f_{2} x y z=x \\
& f_{3}:: \alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha \\
& f_{3} x y z=y \\
& f_{4}:: \alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha \\
& f_{4} x y z=z
\end{aligned}
$$

This observation can also be considered as the basis for the validity of free theorems. In other words, for a function with a polymorphic type there are only a couple of semantically distinguishable implementations. And, therefore, we can check whether
two functions of type $[\alpha] \rightarrow[\alpha]$ are equal by only checking them for a small number of arguments, namely, all images of shape.

We can regard a function of type $\alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$ as a function of type $[\alpha] \rightarrow[\alpha]$ that takes lists with exactly three elements and yields lists with exactly one element. As Theorem 6.3.2 states, two functions of this kind can be distinguish by checking their behavior for the argument $[0,1,2]$. In other words, to check whether two functions of type $\alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$ are equal we only have to apply them to the arguments 0,1 , and 2 . If we consider the definitions above, this becomes quite obvious. That is, for a function of type $f:: \alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$ we can check whether it is equivalent to $f_{1}$, $f_{2}, f_{3}$, or $f_{4}$ by checking whether the application $f 012$ yields $\perp, 0,1$, or 2 .

Now let us consider a function of type $\alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$ in the presence of seq. If we consider a function $f:: \alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$, this function has to be sequential if we consider a sequential language. However, free theorems do not consider sequentiality. For example, by the corresponding free theorem, $f$ might satisfy f $x \perp \perp \equiv x$, $f \perp x \perp \equiv x$, and $f \perp \perp x \equiv x$ for $x \not \equiv \perp$. Though, the function $f$ would not be sequential as no argument is a sequential position.

In a sequential language we have either $f \perp y z \equiv \perp, f x \perp z \equiv \perp$, or $f x y \perp \equiv \perp$ for all $x, y$, and $z$. By sequentiality at least two of the applications $f x \perp \perp, f \perp x \perp$, or $f \perp \perp x$ yield $\perp$ for all $x$ as one of the arguments is a sequential position. Without loss of generality we assume that there exists an $x$ such that the application $f x \perp \perp$ does not yield $\perp$. A free theorem states that we have $f x \perp \perp \equiv x$ in this case. Therefore, regarding a function $f:: \alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$, the behavior of the applications $f x \perp \perp, f \perp x \perp$, and $f \perp \perp x$ for all $x$ is already determined by sequentiality. Thus, we do not have to consider the corresponding test cases.

If we consider a function of type $\alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$ as a function of type $[\alpha] \rightarrow[\alpha]$, again, we would check it for the images of shape seq , that is, for the lists

$$
[\perp, \perp, \perp],[0, \perp, \perp],[\perp, 1, \perp],[\perp, \perp, 2],[\perp, 1,2],[0, \perp, 2],[0,1, \perp] \text {, and }[0,1,2] .
$$

However, there are no two functions of type $\alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$ that can be distinguished by the test cases $f \perp \perp \perp, f 0 \perp \perp, f \perp 1 \perp$, or $f \perp \perp 2$ and cannot be distinguished by one of the other test cases. Thus, using the test cases $f 01 \perp$, $f 0 \perp 2, f \perp 12$, and $f 012$ we can already distinguish all functions of type $\alpha \rightarrow \alpha \rightarrow \alpha \rightarrow \alpha$.

We assume that we can generalize this observation from functions of type $\alpha \rightarrow$ $\alpha \rightarrow \alpha \rightarrow \alpha$ to functions on type $f::[\alpha] \rightarrow[\alpha]$ in a setting with seq primitive. Instead of checking functions of type $[\alpha] \rightarrow[\alpha]$ for all images of shape ${ }_{\text {seq }}$ we can reduce the number of test cases. We have to check functions of type $[\alpha] \rightarrow[\alpha]$ only for all shapes where we replace one element of the list with $\perp$. That is, we do not have to consider test cases like $[0, \perp, \perp]$ or $\perp: 1: \perp: 2: \perp$. Note, that this reduces the number of test cases from exponentially many in the length of the list to linearly many in the length of the list. However, note that this observation does not hold in the presence of a feature like unamb :: $\alpha \rightarrow \alpha \rightarrow \alpha$ (Elliott 2009) as it allows to define non-sequential functions. Refining the concept of a shape in the presence of seq as illustrated above is future work. More precisely, we would have to incorporate sequentiality into the logical relation that is the basis for free theorems.

### 6.5 Minimally Strict Functions on Lists

We return to a setting without seq to consider minimally strict functions. If we reconsider Theorem 6.3.2, one might assume that we can check whether a polymorphic function $f::[\alpha] \rightarrow[\alpha]$ is minimally strict by checking whether there exists a function $h::[$ Int $] \rightarrow[$ Int $]$ that is less strict than $f_{\text {Int }}$ with respect to all shapes. Consider the following function of type $[\alpha] \rightarrow[\alpha]$ that yields a singleton list that contains the first element of its argument list.

$$
\begin{aligned}
& f::[\alpha] \rightarrow[\alpha] \\
& f(x:-)=[x]
\end{aligned}
$$

The following function $h::[$ Int $] \rightarrow[$ Int $]$ is less strict than $f_{\text {Int }}$ with respect to all shapes as we have $h \perp \equiv[0] \sqsupset \perp \equiv f_{\text {Int }} \perp$, for example.

$$
\begin{aligned}
& h::[\text { Int }] \rightarrow[\text { Int }] \\
& h_{-}=[0]
\end{aligned}
$$

Though, obviously there exists no polymorphic function whose integer instance behaves like $h$. The function $h$ "invents" a list element, namely, 0. As Lemma 6.3.5 tells us, the elements of the result list of a polymorphic function of type $[\alpha] \rightarrow$ $[\alpha]$ are taken from the argument list or they are $\perp$. To prevent the monomorphic function from inventing elements we do not compare $f_{\text {Int }} x s$ with $h x s$ but with map (xs!!) ( $h x s$ ). In the example above, $h \perp$ is more defined than $f_{\text {Int }} \perp$, but map $(\perp!!)(h \perp)$ is as defined as $f_{\text {Int }} \perp$. Therefore, $h$ is not a witness that shows that $f$ is unnecessarily strict.

If we consider a less-strict relation that only relates functions if they agree for total arguments, then $h$ is not less strict than $f_{\text {Int }}$ as we have $h[] \equiv[0]$, but $f_{\text {Int }}[] \equiv \perp$. Note that we cannot resolve this issue by adding a rule for the empty list to $f$ as it would have to invent a list element. However, even in a setting that only relates functions by the less-strict relation if they agree for total arguments, we can define a similar example by considering the following data type of non-empty lists.

$$
\text { data NonEmpty } \alpha=\text { NonEmpty } \alpha[\alpha]
$$

We have to define a notion of shape and indexing for the data type NonEmpty, which is straightforward. Then, we can define a polymorphic function similar to $f$ and a monomorphic function similar to $h$ such that $h$ is less strict than $f_{\text {Int }}$ but $f$ is still minimally strict. Furthermore, $h$ agrees with $f_{\text {Int }}$ for all total arguments.
Theorem 6.5.1: Let $f::[\alpha] \rightarrow[\alpha]$. The function $f$ is not minimally strict if and only if there exists a function $h::[$ Int $] \rightarrow[$ Int $]$ such that

$$
g \prec f
$$

where $g$ is defined as follows.

$$
\begin{aligned}
& g::[\alpha] \rightarrow[\alpha] \\
& g x s=\operatorname{map}(x s!!)(h(\text { shape } x s))
\end{aligned}
$$

Proof: $\Longrightarrow$ : As $f$ is not minimally strict, there exists a polymorphic function $g::$ $[\alpha] \rightarrow[\alpha]$ that is less strict than $f$. By Lemma 6.3.4 we have

$$
g x s \equiv \operatorname{map}(x s!!)\left(g_{\text {Int }}(\text { shape } x s)\right)
$$

and take $g_{\text {Int }}$ as $h$.
$\Longleftarrow$ : The function $f$ is not minimally strict since $g::[\alpha] \rightarrow[\alpha]$ is less strict than $f$.

We can prove a similar theorem in the presence of seq by employing shape ${ }_{\text {seq }}$ instead of shape.

Sloth uses a slightly different approach to check whether a polymorphic function is minimally strict. Instead of considering a polymorphic function $g$ as defined in Theorem 6.5.1, with respect to a function $h::$ [Int] $\rightarrow$ [Int] Sloth considers a monomorphic function $g x s=$ map ( $x s!!$ ) ( $h x s$ ). To check whether a polymorphic function $f$ is minimally strict we check whether we have $g \circ$ shape $\prec f_{\text {Int }} \circ$ shape. It is easy to prove this refined statement by Theorem 6.5.1 and Theorem 6.3.2.

By employing the standard approach for monomorphic functions we can check whether for a polymorphic function $f::[\alpha] \rightarrow[\alpha]$ there exists a function $h::[$ Int $] \rightarrow$ [Int] that is less strict than $f_{\text {Int }}$ with respect to some shape. To consider shapes only, the test case generator for the type $[A]$ simply generates list test cases as usually but replaces occurrences of elements of type $A$ by the index of the element in the list. After we have discovered a shape ys $::[$ Int $]$ such that $h$ ys $\sqsupset f$ ys, we check whether we also have map (ys!!) (hys) $\sqsupset f$ ys. If this is the case, then $y s$ is actually a counter-example, and we propose map (ys!!) (hys) as result of $f$ for the argument $y s$. Sloth presents the counter-example $y s$ by substituting indices by unique names.

Note that we cannot check whether a polymorphic function that uses seq is minimally strict by using the $A$-instance as we use the notion of shape as defined by shape and not by shape ${ }_{\text {seq }}$.

### 6.6 Generalization

In this section we consider polymorphic functions that do not have the type $[\alpha] \rightarrow$ $[\alpha]$. For example, consider the following data type for binary trees.

$$
\text { data Tree } \alpha=\text { Empty } \mid \text { Node }(\text { Tree } \alpha) \alpha(\text { Tree } \alpha)
$$

We define a function breadthFirst that enumerates the elements of a tree in breath-first order. The function partition takes a list of values and a list of subtrees and adds the components of a Node to the corresponding lists.

$$
\begin{aligned}
& \text { partition :: Tree } \alpha \rightarrow([\alpha],[\text { Tree } \alpha]) \rightarrow([\alpha],[\text { Tree } \alpha]) \\
& \text { partition Empty level }=\text { level } \\
& \text { partition }(\text { Node } l v r)(v s, \text { trees })=(v: v s, l: r: \text { trees })
\end{aligned}
$$

The function breadthLevel takes one level of a tree, splits the level into values and the next level by employing partition and concatenates the values of the current level with the recursive result.

```
breadthLevel :: \([\) Tree \(\alpha] \rightarrow[\alpha]\)
breadthLevel [] = []
breadthLevel level \(=\) values + breadthLevel nextLevel
    where
        (values, nextLevel) \(=\) foldr partition ([], []) level
breadthFirst :: Tree \(\alpha \rightarrow[\alpha]\)
breadthFirst tree \(=\) breadthLevel \([\) tree \(]\)
```

We can check whether breadthFirst is minimally strict by instantiating all type variables with the data type $A$.

```
> strictCheck (breadthFirst :: Tree A -> [A]) 3
5: \(Node (Node }\perp\textrm{a}\perp\mathrm{ ) b L) -> b:a: }
Finished 9 tests.
```

To investigate why breadthFirst is unnecessarily strict we check whether partition is minimally strict. Sloth reports the following counter-examples.

```
> strictCheck
    (partition :: Tree A -> ([A],[Tree A]) -> ([A],[Tree A])) 3
1: \\perp \perp -> ( }\perp,\perp
2: \Empty \perp -> ( }\perp,\perp
3: \(Node }\perp\textrm{a} \perp) \perp -> (a: , , \perp: \perp: \perp) 
Finished 5 tests.
```

The first counter-example states that partition applied to $\perp$ and $\perp$ yields $\perp$ while there exists is less strict implementation that yields $(\perp, \perp)$ instead. And, even if the first argument is not $\perp$ the function still yields $\perp$ as result. As partition performs pattern matching on the tuple in the second argument, as long as the second argument is $\perp$, the result is $\perp$ as well. Although this looks like a minor problem and will not lead to bad performance in most cases, the consequences for breadthFirst are heavy. The implementation presented above does not yield any value of a certain level if any of the trees on the same level is $\perp$. In other words, to yield the first value of a level we have to evaluate all trees of this level to head normal form. Note that these are exponentially many trees in the depth of the tree on a particular level.

In order to improve the implementation we simply have to replace the tuple pattern matching in partition by a lazy pattern matching. This minor change has a quite significant effect on the runtime of the function. Furthermore, this performance gain increases with larger tree sizes.

To illustrate the benefit of the less strict implementation we perform a breadth first traversal of a complete binary tree and calculate the length of the resulting list. The following table presents the run-times for the presented implementation of breadthFirst and a less strict implementation with a lazy pattern matching, called breadthFirst'.

| depth | breadthFirst | breadthFirst $^{\prime}$ |
| :---: | ---: | ---: |
| 20 | 2.61 s | 0.78 s |
| 21 | 8.67 s | 1.65 s |
| 22 | 30.91 s | 3.16 s |

These times were measured with an initial stack size of 50 megabytes as the unnecessarily strict implementation runs out of stack otherwise.

To test whether breadthFirst is minimally strict, Sloth uses a generalization of the results presented in the previous sections. More precisely, Sloth, for example, only uses the following test cases to check breadthFirst for trees up to size three.

$$
\perp, \text { Empty, Node } \perp a \perp, \text { Node Empty } a \perp, \text { Node }(\text { Node } \perp a \perp) b \perp
$$

The values $a$ and $b$ can be considered as integers that uniquely identify the position of these values in the term. Sloth does not check the test cases Node $\perp$ a Empty as well as Node $\perp a$ (Node $\perp b \perp$ ) because the last argument of Node is not a sequential position in Node $\perp a \perp$ with respect to breadthFirst and, therefore, these values are not elements of the characteristic set. In this section we sketch the idea that allows us to reduce the number of test case for breadthFirst to a linear number of test cases in the number of elements in the tree. More details about this approach and the corresponding proofs for the generalization can be found in (Christiansen 2011).

Instead of considering functions of type $[\alpha] \rightarrow[\alpha]$ only, we consider functions of type $\varphi \alpha \rightarrow \psi \alpha$ where $\varphi$ and $\psi$ are functors that are isomorphic to a functor composed of the unit, the identity, the constant, the product, and the sum functor. In the following, we refer to functors constructed from these building blocks as generic functors. The approach to generic functors is very similar to the approach by Magalhães et al. (2010). For arbitrary functors $\varphi$ the user has to provide instances of overloaded functions from and to, which constitute an embedding projection pair. The function from transforms a value of type $\varphi \tau$ into a value of type Gen $\varphi \tau$, where Gen is a function on the type level (Chakravarty et al. 2005) that maps a type constructor to the isomorphic generic functor. In other words, Gen $\varphi$ denotes a functor that is a generic representation of $\varphi$. In the same way, the function to maps generic values, that is, values of type Gen $\varphi \tau$ to elements of the original type, namely, elements of type $\varphi \tau$.

To generalize the statements from the previous sections we have to provide generalizations of map, (!!), and shape for arbitrary first-order functors. Therefore, we define instances of the type class Functor for all generic functors. Furthermore, we define two type classes that provide functions called gproj and gshape and define instances of these type classes for all generic functors. By means of a structural induction over the structure of generic functors we can then prove the following statement.
Lemma 6.6.1: For all generic functors $\varphi$ and all $x:: \varphi \tau$ the following holds.

$$
\text { fmap }(\text { gproj } x)(\text { gshape } x) \equiv x
$$

By employing the functions from and to and the instances of fmap, gproj, and gshape for generic functors we get instances of the functions for all functors that can be
represented by generic functors. This way we can easily generalize the previous lemma to all functors that provide an embedding projection pair from and to. To generalize all the other statements from functions of type $[\alpha] \rightarrow[\alpha]$ to functions of type $\varphi \alpha \rightarrow \psi \alpha$ we basically have to replace occurrences of map, (!!), and shape in the proofs by fmap, gproj, and gshape.

## 7 Case Studies

In this chapter we present several case studies of applications of Sloth. Before we apply Sloth to check whether several functions are minimally strict, in Section 7.1 we show how we can, in some cases, derive a less strict implementation. In Section 3.2 we have already used this technique to derive a less strict implementation of intersperse, and in Section 6.1 we have used this technique to derive a less strict implementation of inits.

When we have the means to derive less strict implementations at hand, in Section 7.2 we check whether the standard implementation of the multiplication of Peano numbers is minimally strict. This function is a simple monomorphic example of an unnecessarily strict function that furthermore demonstrates that we can benefit from less strict function definitions if we consider infinite data structures.
In Section 7.3 we consider an implementation of binary numbers as algebraic data type by Braßel et al. (2008). This example is of particular interest as it demonstrates that minimally strict functions provide additional benefits in a functional logic programming language like Curry (Hanus 2006).

In Section 7.4 we consider a somehow more practical example. We examine the implementation of a function from the Haskell library split, which provides a variety of functions to split a list into sublists.

Finally, in Section 7.5 we show that minmal strictness cannot be put on a level with low memory footprint or even fast execution. More precisely, we consider a function that reverses a list and show that the benefit of a minimally strict implementation highly depends on the context of an application of a minimally strict function.

### 7.1 Deriving a Less Strict Implementation

As we have observed in Section 3.2, in some cases we can derive a less strict implementation of a function by distributing a context that occurs on all right-hand sides of a case expression over the case expression. That is, we consider a function $f$ of the following form where $C$ and $D$ are contexts.

$$
\begin{aligned}
f x_{1} \ldots x_{n}=C[\text { case } e & \text { of } \\
p_{1} & \rightarrow D\left[e_{1}\right] \\
& \ldots \\
p_{m} & \left.\rightarrow D\left[e_{m}\right]\right]
\end{aligned}
$$

Here, a context is an expression with the additional symbol [•], called hole. By C[e] we denote the substitution of the hole by the expression $e$.

If the variables of the patterns $p_{1}$ to $p_{m}$ do not occur free in $D$, we can transform the function $f$ into a function $g$.

$$
\begin{aligned}
g x_{1} \ldots x_{n}=C[D[\text { case } e & \text { of } \\
p_{1} & \rightarrow e_{1} \\
& \cdots \\
p_{m} & \left.\left.\rightarrow e_{m}\right]\right]
\end{aligned}
$$

In the following we refer to the transformation from the former into the latter form as case deferment.

The function $g$ is at least as little strict as the function $f$. Though, in general, $g$ is not necessarily less strict than $f$. To determine the conditions to get a less strict implementation by case deferment, we consider this transformation in the simple functional programming language introduced in Section 2.2. As this programming language does not provide arbitrary algebraic data types, we examine the special case of a case expression that matches lists. That is, we have a function of the following form where $y$ and $y s$ do not occur free in the context $D$.

$$
\left.\begin{array}{rl}
f x=C[\text { case } e \text { of } & \\
& \rightarrow D\left[e_{1}\right] \\
& \text { Cons }\langle y, y s\rangle
\end{array} \rightarrow D\left[e_{2}\right]\right] .
$$

We calculate the semantics of $f$ with respect to the semantics defined in Figure 2.2.4. The semantics of a context $C$, denoted by $\llbracket C \rrbracket_{a}$ where $a$ is an environment, is a function that takes a semantic value $v$, fills the hole in the context with a fresh variable $x$ and updates the environment such that occurrences of $x$ are replaced by $v$. More precisely, we define

$$
\llbracket C \rrbracket_{a}=\lambda v \cdot \llbracket C[x] \rrbracket_{a[x \mapsto v]} .
$$

To calculate the semantics of $f$ we use the fact that the semantics is compositional, in other words, we use the equation

$$
\llbracket C[e] \rrbracket_{a}=\llbracket C \rrbracket_{a} \llbracket e \rrbracket_{a} .
$$

In the following we abbreviate the environment $a[x \mapsto v]$ to $a^{\prime}$ and calculate the semantics of $f$.

$$
\begin{aligned}
\llbracket f \rrbracket_{a} & =\lambda v \cdot \llbracket C\left[\text { case } e \text { of }\left\{N i l \rightarrow D\left[e_{1}\right] ; \text { Cons }\langle y, y s\rangle \rightarrow D\left[e_{2}\right]\right\}\right] \rrbracket_{a^{\prime}} \\
& =\lambda v \cdot \llbracket C \rrbracket_{a^{\prime}} \begin{cases}\perp & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\perp \\
\llbracket D\left[e_{1}\right] \rrbracket_{a^{\prime}} & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Nil } \\
\llbracket D\left[e_{2}\right] \rrbracket_{a^{\prime}\left[\left\{y \mapsto v_{1}, y s \mapsto v_{2}\right]\right.} & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle\end{cases} \\
& =\lambda v \cdot \llbracket C \rrbracket_{a^{\prime}} \begin{cases}\perp & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\perp \\
\llbracket D \rrbracket_{a^{\prime}} \llbracket e_{1} \rrbracket_{a^{\prime}} & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Nil } \\
\llbracket D \rrbracket_{a^{\prime}\left[y \mapsto v_{1}, y s \mapsto v\right]} \llbracket e_{2} \rrbracket_{a^{\prime}\left[y \mapsto v_{1}, y s \mapsto v_{2}\right]} & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle\end{cases} \\
& =\lambda v \cdot \llbracket C \rrbracket_{a^{\prime}} \begin{cases}\perp & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\perp \\
\llbracket D \rrbracket_{a^{\prime}} \llbracket e_{1} \rrbracket_{a^{\prime}} & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Nil } \\
\llbracket D \rrbracket_{a^{\prime}} \llbracket e_{2} \rrbracket_{a^{\prime}\left[y \mapsto v_{1}, y s \mapsto v_{2}\right]} & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle\end{cases}
\end{aligned}
$$

$$
=\lambda v \cdot \begin{cases}\llbracket C \rrbracket_{a^{\prime}} \perp & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\perp \\ \llbracket C \rrbracket_{a^{\prime}}\left[\llbracket D \rrbracket_{a^{\prime}} \llbracket e_{1} \rrbracket_{a^{\prime}}\right) & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Nil } \\ \llbracket C \rrbracket_{a^{\prime}}\left(\llbracket D \rrbracket_{a^{\prime}} \llbracket e_{2} \rrbracket_{a^{\prime}\left[y \mapsto v_{1}, y s \mapsto v_{2}\right]}\right) & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle\end{cases}
$$

The last but one step of the calculation is valid because we only consider contexts $D$ that do not contain free occurrences of $y$ and $y s$. In the same way we calculate the semantics of the corresponding function $g$.

$$
\begin{aligned}
\llbracket g \rrbracket_{a} & =\lambda v \cdot \llbracket C\left[D\left[\text { case } e \text { of }\left\{N i l \rightarrow e_{1} ; C o n s\langle y, y s\rangle \rightarrow e_{2}\right\}\right\rfloor\right] \rrbracket_{a^{\prime}} \\
& =\lambda v \cdot \llbracket C \rrbracket_{a^{\prime}}\left(\begin{array}{ll}
\llbracket D \rrbracket_{a^{\prime}} & \left.\begin{array}{ll}
\perp & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\perp \\
\llbracket e_{1} \rrbracket_{a^{\prime}} & \text { if } \llbracket e \rrbracket_{a^{\prime}}
\end{array}\right) \\
\llbracket e_{2} \rrbracket_{a^{\prime}}\left[y \mapsto v_{1}, y s \mapsto v_{2}\right] & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle
\end{array}\right) \\
& =\lambda v \cdot \begin{cases}\llbracket C \rrbracket_{a^{\prime}}\left(\llbracket D \rrbracket_{a^{\prime}} \perp\right) & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\perp \\
\llbracket C \rrbracket_{a^{\prime}}\left(\llbracket D \rrbracket_{a^{\prime}} \llbracket e_{1} \rrbracket_{a^{\prime}}\right) & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Nil } \\
\llbracket C \rrbracket_{a^{\prime}}\left(\llbracket D \rrbracket_{a^{\prime}} \llbracket e_{2} \rrbracket_{a^{\prime}\left[y \mapsto v_{1}, y s \mapsto v_{2}\right]}\right) & \text { if } \llbracket e \rrbracket_{a^{\prime}}=\text { Cons }\left\langle v_{1}, v_{2}\right\rangle\end{cases}
\end{aligned}
$$

Thus, if we consider functions $g$ and $f$ of type $\tau_{1} \rightarrow \tau_{2}$ we have $\llbracket g \rrbracket \sqsupset \llbracket f \rrbracket$ if there exists $v \in \llbracket \tau_{1} \rrbracket$ such that

$$
\llbracket e \rrbracket_{[x \mapsto v]}=\perp
$$

and

$$
\llbracket C \rrbracket_{[x \mapsto v]}\left(\llbracket D \rrbracket_{[x \mapsto v]} \perp\right) \sqsupset \llbracket C \rrbracket_{[x \mapsto v]} \perp .
$$

As case deferment is a purely syntactical transformation, a compiler could apply this transformation automatically by inspecting the right-hand sides of a case expression. In some cases, we get a less strict implementation by applying this simple transformation. However, even if we do not get a less strict function, the resulting function is as little strict as the original one. Furthermore, Gustavsson and Sands (1999) show that case deferment preserves heap usage and improves stack usage if $D$ is a context that evaluates the hole to head normal form if the context itself is evaluated to head normal form. They refer to this kind of contexts as reduction contexts. For all reduction contexts $D$ we have $\llbracket D \rrbracket_{[x \mapsto v]} \perp=\perp$ and, therefore, $\llbracket C \rrbracket_{[x \mapsto v]}\left(\llbracket D \rrbracket_{[x \mapsto v]} \perp\right)=\llbracket C \rrbracket_{\{x \mapsto v]} \perp$. Thus, they consider a form of case deferment that does not yield a function that less strict. Note that it is not a necessary condition for $\llbracket C \rrbracket_{[x \mapsto v]}\left(\llbracket D \rrbracket_{[x \mapsto v]} \perp\right)=\llbracket C \rrbracket_{[x \mapsto v]} \perp$ that $D$ is a reduction context. They do not consider the more general transformation presented here as it is not semantics preserving. We did not check whether, using the theory by Gustavsson and Sands (1999, 2001), we can prove that case deferment never yields a function that consumes more memory than the original implementation. However, a syntactical approach to less strict implementations is certainly a direction for future considerations.

In the following sections, we will see some examples of functions that can be improved with respect to strictness by applying case deferment. However, in most cases, we have to apply other transformations first before we can apply case deferment. Furthermore, we will as well see several examples where case deferment

## 7 Case Studies

cannot be used to derive a less strict implementation. In these cases we have to alter the order of pattern matching.

The following lemma summarizes the observations so far, but, in contrast to the considerations above, we do not use our simple functional language but full Haskell and arbitrary algebraic data types instead of lists. Furthermore, we use an informal notation where we do not separate syntax and semantics in the same line as we do in Chapter 6. We do not present a proof for this statement as we would have to define a semantics that provides arbitrary algebraic data types.

Lemma 7.1.1: Let $f, g:: \tau_{1} \rightarrow \cdots \rightarrow \tau_{n} \rightarrow \tau$ be functions defined as follows where $C$ and $D$ are contexts of appropriate types, $p_{1}, \ldots, p_{m}$ are patterns of appropriate types, and $e_{1}, \ldots$, $e_{m}$ and $e$ are expressions of appropriate types. Furthermore, the variables of $p_{1}, \ldots, p_{m}$ do not occur free in the context $D$.

$$
\begin{aligned}
f x_{1} \ldots x_{n}=C[\text { case } e & \text { of } \\
p_{1} & \rightarrow D\left[e_{1}\right] \\
& \ldots \\
p_{m} & \left.\rightarrow D\left[e_{m}\right]\right] \\
g x_{1} \ldots x_{n}=C[D[\text { case } e & \text { of } \\
p_{1} & \rightarrow e_{1} \\
& \ldots \\
p_{m} & \left.\left.\rightarrow e_{m}\right]\right]
\end{aligned}
$$

We have $g \sqsupset f$ if and only if there exist arguments $v_{1}:: \tau_{1}, \ldots, v_{n}:: \tau_{n}$ such that

$$
e_{\left[x_{1} \mapsto v_{1}, \ldots, x_{n} \mapsto v_{n}\right]} \equiv \perp
$$

and

$$
D[C[\perp]]_{\left[x_{1} \mapsto v_{1}, \ldots, x_{n} \mapsto v_{n}\right]} \sqsupset D[\perp]_{\left[x_{1} \mapsto v_{1}, \ldots, x_{n} \mapsto v_{n}\right]} .
$$

Here, $e_{\left[x_{1} \mapsto v_{1}, \ldots, x_{n} \mapsto v_{n}\right]}$ denotes the semantics of e with respect to the given substitutions.
Let us consider two examples of functions that do not become less strict by applying case deferment. Consider the following functions of type $\tau \rightarrow[\tau]$.

$$
\begin{aligned}
& f x=\text { case }[] \text { of } \\
& p_{1} \rightarrow x: e_{1} \\
& \ldots \\
& p_{m} \rightarrow x: e_{m} \\
& g x=x: \text { case }[] \text { of } \\
& p_{1} \rightarrow e_{1} \\
& \cdots \\
& p_{m} \rightarrow e_{m}
\end{aligned}
$$

As we have []$\not \equiv \perp$ for all arguments $v:: \tau$, we get $g \equiv f$ by Lemma 7.1.1. As another example, consider the following functions of type $\tau \rightarrow[\tau]$.

$$
\begin{aligned}
& f x=\operatorname{tail}(\text { case } e \text { of } \\
& p_{1} \rightarrow x: e_{1} \\
& \ldots \\
& \left.p_{m} \rightarrow x: e_{m}\right) \\
& g x=\operatorname{tail}(x: \text { case } e \text { of } \\
& p_{1} \rightarrow e_{1} \\
& \cdots \\
& \left.p_{m} \rightarrow e_{m}\right)
\end{aligned}
$$

As we have tail $(v: \perp) \equiv \perp$ for all arguments $v:: \tau$, we get $g \equiv f$ by Lemma 7.1.1. Note that this is a case where the context $D$ is not a reduction context as considered by Gustavsson and Sands (1999) because $x:[\cdot]$ does not evaluate its hole to head normal form if it is evaluated to head normal form itself.

As an example for a successful application of Lemma 7.1.1 we consider the definition of intersperse from Section 3.2. First, we define intersperse by means of the Haskell fixpoint operator fix :: $(\alpha \rightarrow \alpha) \rightarrow \alpha$. That is, we consider the following functional, as we have intersperse $\equiv$ fix body.

```
body \(::(\alpha \rightarrow[\alpha] \rightarrow[\alpha]) \rightarrow \alpha \rightarrow[\alpha] \rightarrow[\alpha]\)
body _ _ [] \(=[]\)
body \(f \operatorname{sep}(x: x s)=\)
    case \(x s\) of
    [] \(\rightarrow x\) : []
    \({ }_{-} \rightarrow x:\) sep : \(f\) sep \(x s\)
```

We apply case deferment to derive the following less strict implementation of the functional body.

$$
\begin{aligned}
& \text { body' }::(\alpha \rightarrow[\alpha] \rightarrow[\alpha]) \rightarrow \alpha \rightarrow[\alpha] \rightarrow[\alpha] \\
& \text { body }:-[]=[] \\
& \text { body } y^{\prime} f \operatorname{sep}(x: x s)= \\
& x: \text { case } x \text { s of } \\
& \quad[] \rightarrow[] \\
& \quad-\rightarrow \text { sep }: f \text { sep } x s
\end{aligned}
$$

To prove that body' is indeed less strict than body, we have to provide arguments $f$, sep, and $x: x s$ such that $x s \equiv \perp$ and $x: \perp \sqsupset \perp$. These requirements are met for an arbitrary function $f:: \tau \rightarrow[\tau] \rightarrow[\tau]$, an arbitrary element sep $:: \tau$ and a list $x: \perp$, where $x$ is an arbitrary value of type $\tau$. In particular this proves that we have body' $f \sqsupset$ body $f$ for all functions $f:: \tau \rightarrow[\tau] \rightarrow[\tau]$. We will later employ this observation to show that fix body' is less strict than fix body.

As we are not interested in a less strict functional but in a less strict recursive definition of intersperse, we have to provide some means to prove that fix body' is indeed less strict than fix body. Note that case deferment can only be applied to non-recursive definitions, as the right-hand sides of the case expressions of the functions $f$
and $g$ have to agree except for the context $D$. In other words, we cannot apply case deferment to functions $f$ and $g$ if the right-hand sides of the case expressions of $f$ and $g$ contain recursive applications of the corresponding functions. Therefore, we prove the following Lemma by means of fixpoint induction (Bakker 1980).

Lemma 7.1.2: Let $f, g:: \tau \rightarrow \tau$. If we have $g x \sqsupset f x$ for all $x:: \tau$, then fix $g \sqsupset$ fix $f$.
Proof: We prove this statement by fixpoint induction. We define a predicate $P(x, y)$ that takes two values of type $\tau$ and is satisfies if and only if $g x \sqsupset f y$. For the base case of the induction we have to prove that $P(\perp, \perp)$ holds. As we have $g x \sqsupset f x$ for all $x:: \tau$ by precondition, the base case holds. For the inductive step we have to show that for all $x, y:: \tau$ if $P(x, y)$ holds, then $P(g x, f y)$ also holds. We reason as follows.

```
g(g x)
    \sqsupset {precondition}
f(gx)
    { induction hypothesis, monotonicity of f }
f(fy)
```

Thus, $P(g x, f y)$ holds and, therefore, the inductive step holds as well. By the fixpoint induction principle this implies $P($ fix $g$, fix $f$ ) or, in other words $g(f i x g) \sqsupset$ $f$ ( ix $f$ ) which is equivalent to fix $g \sqsupset$ fix $f$.

As we have observed before, we have body $f \sqsupset \operatorname{body} f$ for all $f:: \tau \rightarrow[\tau] \rightarrow[\tau]$. Thus, by the previous lemma we also have fix body $\sqsupset$ fix body. This shows that intersperse' from Section 3.2 is less strict than intersperse.

The notion of "less strict" considered above refers to the definition of the less-strict relation from Definition 6.1.1. There a function $f$ is less strict than a function $g$ if we have $f \sqsupseteq g$, that is, $f x \sqsupseteq g x$ for all $x$ of appropriate type. In contrast, according to the original definition of the less-strict relation (Definition 4.1.2) a function $f$ is called less strict than a function $g$ if we have $f \sqsupseteq g$, and $f$ and $g$ agree for total arguments. Therefore, we additionally have to show that two functions agree for total arguments to apply these results to the definition presented in Chapter 4. In most cases we consider functions that yield total results for all total arguments. If a function $g$ yields total results for all total arguments and we have $f \sqsupseteq g$, that is, $f x \sqsupseteq$ $g x$ for all $x$ of appropriate type, then $f$ also yields total results for all total arguments. For example, intersperse yields total results for all total arguments. In other words, by applying Lemma 7.1.1 and Lemma 7.1.2 we already know that intersperse' is less strict than intersperse in the sense of Definition 4.1 .2 because intersperse yields total results for all total arguments. We only have to take additional care if we consider a function that yields a non-total result for a total argument like the division function on integers. All functions considered in the following sections yield total results for all total arguments, so we can apply Lemma 7.1.1 and Lemma 7.1.2 to derive less strict implementations in the sense of Definition 4.1.2.

### 7.2 Peano Multiplication

In this section we observe that the standard implementation of the multiplication of Peano numbers is too strict and present a benefit of less strict functions besides memory usage. Later we show that the standard implementation of an intersection of lists shows the same bad behavior as Peano multiplication.

Consider the following data type of Peano numbers.

$$
\text { data Peano }=\text { Zero } \mid \text { Succ Peano }
$$

We define addition and multiplication of Peano numbers as follows.

```
addP \(::\) Peano \(\rightarrow\) Peano \(\rightarrow\) Peano
addP Zero \(\quad n=n\)
\(\operatorname{addP}(\) Succ \(m) n=\) Succ (addP \(m n\) )
multP \(::\) Peano \(\rightarrow\) Peano \(\rightarrow\) Peano
multP Zero \(\quad-=\) Zero
multP \((\) Succ \(m) n=\operatorname{addP} n(\) multP \(m n)\)
```

Furthermore, we define an infinite Peano number called infinity.

```
infinity :: Peano
infinity = Succ infinity
```

This is a standard implementation of Peano number arithmetic. For example, the numbers package by Augustsson (2009) provides an identical implementation.

The evaluation of multP Zero infinity yields Zero in a non-strict programming language like Haskell. On the contrary, the evaluation of multP infinity Zero does not terminate. So, do all implementations of the multiplication of Peano numbers behave this way? aThat is, can we give an implementation of Peano multiplication in Haskell that yields Zero in both cases (without using parallelism)?

Sloth reports two counter-examples if we check multP up to size six.

```
> strictCheck multP 6
4: \(Succ \perp) Zero -> Zero
7: \(Succ (Succ \perp)) Zero -> Zero
Finished 23 tests.
```

As the counter-examples presented by Sloth are potential counter-examples, we have to verify them. For all total inputs that are more defined than Succ $\perp$ and Zero the function multP yields Zero, which is as defined as the recommended result Zero. Therefore, the counter-examples are definitely counter-examples.

In particular, the counter-examples show that a minimally strict implementation of Peano multiplication only has to evaluate the outermost Succ constructor of its first argument if the second argument is Zero. Therefore, a minimally strict implementation yields Zero for the arguments infinity and Zero. Thus, now we are able

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to answer the question. There is an implementation of the multiplication of Peano numbers that terminates no matter whether it is applied to Zero and infinity or to infinity and Zero.

The evaluation of multP infinity Zero does not terminate because multP is inductively defined over its first argument. Hence, even if the second argument is Zero, the first argument is completely evaluated. Note that it, therefore, takes linear time in the size of $p$ to evaluate the application multP $p$ Zero for any Peano number $p$.

To derive a less strict implementation, we inline the definition of addP into multP and get the following equivalent definition.

```
multP \(::\) Peano \(\rightarrow\) Peano \(\rightarrow\) Peano
multP Zero \(\quad-=\) Zero
multP (Succ m) \(n=\)
    case \(n\) of
        Zero \(\rightarrow\) multP \(m\) Zero
        Succ \(n^{\prime} \rightarrow\) Succ \(\left(\right.\) addP \(m\left(\right.\) multP \(\left(m\left(\right.\right.\) Succ \(\left.\left.\left.\left.n^{\prime}\right)\right)\right)\right)\)
```

If we consider this implementation, we observe that
multP m Zero
in the second rule yields Zero for all Peano numbers $m$. Well, in fact, not for all Peano numbers, if $m$ is terminated by $\perp$, for example, $m \equiv$ Succ (Succ (Succ $\perp$ )), then we have multP $m$ Zero $\equiv \perp$. Therefore, we can replace the application multP $m$ Zero by the constant Zero and get the following less strict implementation of multP. We slightly simplify the definition by employing the fact that pattern matching is performed from top to bottom. However, we still use the pattern (Succ _) instead of _ in the second rule to emphasize that there is a bias towards the first argument with respect to strictness.

$$
\begin{aligned}
& \text { multP } P^{\prime}:: \text { Peano } \rightarrow \text { Peano } \rightarrow \text { Peano } \\
& \text { multP' Zero } \quad=\text { Zero } \\
& \text { mult } P^{\prime} \text { (Succ _) Zero }=\text { Zero } \\
& \text { mult } P^{\prime}(\text { Succ } m) n=\operatorname{addP} n\left(\text { multP }^{\prime} m n\right)
\end{aligned}
$$

Sloth does not report any counter-examples if we check multP ${ }^{\prime}$ for Peano numbers up to size 50. Note that for any Peano number $p$, it takes only constant time to evaluate mult $P^{\prime} p$ Zero rather than linear time as for multP $p$ Zero.

This is an example where we cannot apply case deferment to derive a less strict implementation. Instead we have to change the order of pattern matching. If the first argument has the form Succ $m$, the original implementation performs pattern matching on $m$. In contrast, the less strict implementation performs pattern matching on the second argument if the first argument has the form Succ $m$.

We might doubt the practical relevance of a minimally strict Peano multiplication. Therefore, we consider the function intersect :: Eq $\alpha \Rightarrow[\alpha] \rightarrow[\alpha] \rightarrow[\alpha]$ from the standard Haskell library Data.List that yields the intersection of two lists and is probably of greater practical relevance than Peano multiplication.

In fact, we do not know how we are supposed to check a function that has an equality constraint. We can regard a function with a type class constraint as a high-er-order function as we can apply dictionary transformation (Wadler and Blott 1989). Therefore, we would be able to check intersect if we would be able to handle high-er-order functions. Nevertheless, we conjecture that we can as well instantiate all type variables that have an $E q$ contraint with an integer type to check whether a function is minimally strict.

Although we did not prove this conjecture, we can easily show that it is not sufficient to check a monomorphic instance with respect to a type that has only finitely many values. For example, consider the function nub :: Eq $\alpha \Rightarrow[\alpha] \rightarrow[\alpha]$ that removes duplicate elements from a list. If we check the Boolean instance of $n u b$, Sloth reports the following counter-examples.

```
> strictCheck (nub :: [Bool] -> [Bool]) 7
11: \(True:False:\perp) -> True:False:[]
12: \(False:True:\perp) -> False:True:[]
17: \(True:False: }\perp:\perp) -> True:False:[]
19: \(False:True: }\perp:\perp) -> False:True:[]
24: \(True:False:\perp:[]) -> True:False:[]
26: \(False:True: }\perp:[]) -> False:True:[]
Finished 45 tests.
```

These counter-examples state that, if the first two elements of a list are the Boolean values False and True, the function $n u b$ does not have to inspect more elements as the result will be a list that only contains the elements False and True anyway. For every monomorphic instance with respect to a type that has only finitely many values we get similar counter-examples. This example demonstrates that we cannot use a monomorphic instance with respect to a type that has only finitely many values to check whether a polymorphic function with an $E q$ constraint is minimally strict. Thus, in the following we consider the integer instance of a polymorphic function with $E q$ constraint to check whether the function is minimally strict.

Back on the road we consider the following implementation of intersect. For simplicity, this implementation is an equivalent variation of the actual implementation in Data.List. The function elem :: Eq $\alpha \Rightarrow \alpha \rightarrow[\alpha] \rightarrow$ Bool takes a value and a list and checks whether the value is an element of the list.

$$
\begin{aligned}
& \text { intersect :: } E q \alpha \Rightarrow[\alpha] \rightarrow[\alpha] \rightarrow[\alpha] \\
& \text { intersect }[]-=[] \\
& \text { intersect }(x: x s) \text { ys } \\
& \mid x \text { 'elem' ys }=x: \text { intersect } x s \text { ys } \\
& \mid \text { otherwise }=\text { intersect } x s \text { ys }
\end{aligned}
$$

Note that, if the first argument of intersect contains duplicate elements, then the result list will contain duplicate elements as well.

We check the implementation of the integer instance of intersect for arguments up to size five. Sloth reports the following counter-example. Note that we, in fact, use a data type with a finite number of constructors again as Int is restricted to a

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finite number of bits ${ }^{1}$. Nevertheless, as the number of constructors is quite large, technically, the finiteness of Int does not cause problems.

```
> strictCheck (intersect :: [Int] -> [Int] -> [Int]) 5
4: \( }\perp:\perp) [] -> []
Finished 13 tests.
```

As we have intersect $x s[] \equiv[]$ for all total list $x s::[I n t]$, the potential counter-example is definitely a counter-example. In fact, we do not only have intersect ( $\perp$ : $\perp$ ) [] $\equiv \perp$, as stated by the counter-example, but intersect $x s[] \equiv \perp$ for all lists $x$ s that are terminated by $\perp$. For example, we have

$$
\text { intersect (replicate } n x+\perp \text { ) }[] \equiv \perp
$$

for all $n::$ Int and all values $x:: \tau$. Here, the application replicate $n x$ yields a list with $n$ elements that only contains the value $x$. Furthermore, in the same way as the evaluation of multP $p$ Zero takes linear time in the size of $p$ the evaluation of intersect $x$ s [] takes linear time in the size of $x$ s.

We can define a less strict implementation of intersect by checking whether the second argument is the empty list in the same way as mult $P^{\prime}$ checks whether the second argument is Zero.

$$
\begin{aligned}
& \text { intersect }^{\prime}:: E q \alpha \Rightarrow[\alpha] \rightarrow[\alpha] \rightarrow[\alpha] \\
& \text { intersect }^{\prime}[] \quad-=[] \\
& \text { intersect }^{\prime}(-:-)[]=[] \\
& \text { intersect }(x: x s) \text { ys } \\
& \mid x \text { 'elem' } y s=x \text { : intersect' } x s \text { ys } \\
& \mid \text { otherwise }=\text { intersect }^{\prime} x s \text { ys }
\end{aligned}
$$

Sloth does not report a counter-example if we check this implementation for integer lists up to size ten. Note that the evaluation of intersect' $x$ s [] takes only constant time.

### 7.3 Binary Arithmetics

In this section we want to emphasize that the concept of minimal strictness is of particular interest if we consider functional logic programming languages. A functional logic programming language like Curry (Hanus 2006) can be considered as a functional language with non-determinism and free variables. In the following, we assume that the reader is familiar with the basic concepts of non-determinism and free variables.

Let us consider the algebraic data type for binary natural numbers presented by Braßel et al. (2008) for the functional logic programming language Curry. The syntax of Curry is very similar to the syntax of Haskell and the deterministic subset of Curry resembles Haskell.

[^15]$$
\text { data Nat }=I H i \mid O \text { Nat } \mid I N a t
$$

This data type defines a little-endian representation of unsigned binary numbers without leading zeros. The constructor IHi represents the most significant one-bit of a binary number, while the constructors $O$ and $I$ represent a zero and a one-bit, respectively. For example, the term $O$ (I IHi) represents the decimal number six. Note that, in contrast to Peano, the data type Nat does not include zero. Braßel et al. (2008) furthermore define the following addition and multiplication of binary numbers, where $\operatorname{succ} N$ yields the successor of a binary number.

```
succN :: Nat }->\mathrm{ Nat
succN IHi =O IHi
succN (On)=In
succN (In) =O(succN n)
addN :: Nat }->\mathrm{ Nat }->\mathrm{ Nat
addN IHi n = succN n
addN (O m)IHi = Im
addN (Om) (On) =O (addN m n)
addN (Om) (In) = I (addN m n)
addN (I m) IHi =O (succN m)
addN (I m) (On) = I (addN m n)
addN (I m) (I n) =O(addN (succN m) n)
multN :: Nat }->\mathrm{ Nat }->\mathrm{ Nat
multN IHi n=n
multN (O m) n=O(multN mn)
multN (I m) n=addN (O(multN mn)) n
```

We use Sloth to check whether multN is unnecessarily strict. Note that we, in fact, check the corresponding Haskell implementation of mult $N$ as Sloth is implemented in Haskell. As mult $N$ is deterministic, it behaves identically in Haskell and Curry if we consider deterministic arguments. We check whether mult $N$ is minimally strict for all pairs of binary numbers up to size five.

```
> strictCheck multN 5
32: \(I \perp) (0 IHi) -> O (I \perp)
33: \(I \perp) (0 (0 \perp)) -> O (0 \perp)
34: \(I \perp) (0 (I \perp)) -> O (I \perp)
Finished 103 tests.
```

Sloth reports three counter-examples. The first counter-example states that, mult $N$ applied to $I \perp$ and $O$ IHi yields $O \perp$ while there exists a less strict implementation that yields $O(I \perp)$ instead.

As all counter-examples consider arguments of the form $I \perp$ and $O m$ for some natural number $m$, we consider the evaluation of the application multN $(I \perp)(O m)$.

```
multN \((I \perp)(O m)\)
    \(\equiv\{\) definition of multN \(\}\)
\(\operatorname{addN}(O(m u l t N \perp(O m)))(O m)\)
    \(\equiv \quad\{\) definition of \(a d d N\}\)
\(O(\) add \(N(m u l t N \perp(O m)) m)\)
    \(\equiv \quad\{\) definition of mult \(N\}\)
\(O(\) add \(N \perp m)\)
    \(\equiv\{\) definition of \(a d d N\}\)
\(O \perp\)
```

If we consider the step labeled with $(*)$, intuitively, we might think that mult $N$ can yield a more defined result. More precisely, for all total natural numbers $m$ and $n$ we have

$$
\text { mult } N n(O m) \equiv O(\operatorname{mult} N n m) .
$$

That is, instead of replacing mult $N \perp(O m)$ by $\perp$ we get a less strict implementation if we replace the application by $O$ (mult $N \perp m$ ). In other words, in the third rule of mult $N$ instead of performing pattern matching on $m$ we would like to perform pattern matching on $n$, that is, the second argument of multN. We get the desired behavior if we swap the arguments of the recursive application of mult $N$ in the third rule of $m u l t N$. That is, we define the following function $m u l t N^{\prime}$.

```
mult \(N^{\prime}::\) Nat \(\rightarrow\) Nat \(\rightarrow\) Nat
mult \(N^{\prime}\) IHi \(\quad n=n\)
mult \(N^{\prime}(O m) n=O\left(m u l t N^{\prime} m n\right)\)
mult \(N^{\prime}(I m) n=a d d N\left(O\left(m u l t N^{\prime} n m\right)\right) n\)
```

We might assume, that this change improves the implementation with respect to some arguments like $I \perp$ and $O I H i$ but worsens it with respect to other arguments. For example, what about the application mult $N(I(O \perp)) \perp$ ? In this case it seems to be disadvantageous to swap the arguments of the recursive application of multN as we get the application mult $N \perp(O \perp)$ instead of the application mult $N(O \perp) \perp$.

Let us consider the application mult $N(I \mathrm{~m}) \perp$ for an arbitrary binary natural number $m$. We get the following equivalence.

```
multN \((\mathrm{Im}) \perp\)
    \(\equiv \quad\{\) definition of multN \(\}\)
addN \((O(\) mult \(N m \perp)) \perp\)
    \(\equiv \quad\{\) definition of \(a d d N\}\)
\(\perp\)
```

As addN is strict in both arguments, no matter what the application of mult $N$ yields, the result of this application is $\perp$ as the second argument of the application of addN is $\perp$ as well. This examples shows that we have

$$
\text { mult } N(\operatorname{I~m}) \perp \equiv m u l t N^{\prime}(\operatorname{I~m}) \perp
$$

for all natural numbers $m$. In other words, because addN consumes both its arguments uniformily it is advantageous if the recursive application of multN in the third rule of multN demands its second argument.

We might still not be convinced that mult $N^{\prime}$ is an improvement with respect to strictness. Thus, we can check whether $m u l t N^{\prime}$ is minimally strict using Sloth. Sloth does not report any counter-example if we check mult $N^{\prime}$ for binary numbers up to size 10. However, note that this only states that $m u l t N^{\prime}$ is probably minimally strict and not that mult $N^{\prime}$ is less strict than mult $N$. The functions mult $N$ and mult $N^{\prime}$ may still be incomparable.

So, why do we bother whether multN is unnecessarily strict? Braßel et al. (2008) have introduced the data type Nat to guess numbers. In a functional logic programming language we can non-deterministically guess elements of a specific type by employing free variables. However, no Curry system can guess values of the primitive integer type. By employing the data type Nat it is possible to guess numbers while arithmetic operations for Nat are still reasonably fast (for example, in comparison to a Peano representation).

In general, in a non-deterministic context, functions that are too strict can lead to unnecessarily many non-deterministic branches, which in turn leads to unnecessary overhead. For example, let us consider the following function.

```
context :: Nat }->\mathrm{ Bool
context (O(I_)) = True
```

The essence of an unnecessarily strict function is that it inspects a larger part of its argument than necessary to yield a certain result. For example, if we evaluate the expression context (multN (I unknown) (O IHi) the function multN performs pattern matching on unknown, where unknown denotes a free variable. Therefore, the free variable is instantiated. When a variable of type Nat is instantiated, it is instantiated to all constructors of its type, applied to fresh free variables, namely, $\mathrm{IHi}, \mathrm{O}$ unknown and I unknown. For all these instantiations the considered expression yields True. Hence, the evaluation yields the result True three times.

```
> context (multN (I unknown) (O IHi))
True
More?
True
More?
True
More?
No more Solutions
```

On the other hand, if we use the less strict implementation mult $N^{\prime}$ we only get a single result.

```
> context (multN' (I unknown) (O IHi))
True
More?
```

No more Solutions

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Here, mult $N^{\prime}$ does not perform pattern matching on unknown. Therefore, the free variable is not instantiated, and the expression is simply evaluated to True.

Thus, minimally strict functions are advantageous in a functional logic programming language. We can observe this advantage when we compare run-times for the generation of Pythagorean triples, using $m u l t N$ and $m u l t N^{\prime}$ respectively. ${ }^{2}$

```
pythagorean :: (Nat,Nat,Nat)
pythagorean \(\mid\) addN \((\) mult \(N a \operatorname{a})(\) mult \(N b b)==\operatorname{mult} N c c=(a, b, c)\)
    where
        \(a, b, c\) free
```

The variables $a, b$, and $c$ are defined to be free variables by the keyword free. That is, pythagorean yields all triples of binary numbers $a, b$, and $c$ such that $a * a+b * b=c * c$. For all numbers that do not satisfy this equation the guard yields a failure, that is, the empty result set.

| number of triples | multN | mult $\mathrm{N}^{\prime}$ |
| :---: | ---: | ---: |
| 100 | 3.87 s | 1.10 s |
| 200 | 17.81 s | 5.87 s |
| 300 | 33.42 s | 12.78 s |

It takes around three times as long to enumerate Pythagorean triples if we use the unnecessarily strict implementation of multiplication.

Most of the functions defined by Braßel et al. (2008) are unnecessarily strict. In the following we consider one additional example, namely, the function cmpNat , which compares two natural numbers and yields a value of type Ordering. The type Ordering has three values, namely, $E Q, L T$, and $G T$, which specify the relation of two values with respect to a given ordering. The following definition of cmpNat employs a function isEQ whose implementation is self-explanatory.

```
compNat \(::\) Nat \(\rightarrow\) Nat \(\rightarrow\) Ordering
compNat IHi IHi =EQ
compNat IHi \(\quad\left(O_{-}\right)=L T\)
compNat \(\mathrm{IHi} \quad\left(I_{-}\right)=L T\)
compNat (O_) IHi =GT
compNat \((O m)(O n)=\mathrm{compNat} m n\)
compNat \((\mathrm{Om})(\mathrm{In})=\)
    if isEQ cmpmn then \(L T\) else cmpmn
        where
            cmpmn \(=\) compNat \(m n\)
compNat (I_) IHi = GT
compNat ( Im ) \((\mathrm{On})=\)
    if isEQ cmpmn then GT else cmpmn
        where
            стрти \(=\) compNat \(m n\)
compNat (I m) (In) = compNat mn
```

[^16]In Haskell we would probably merge the second and the third rule of compNat to a single rule. In Curry these definitions are not equivalent as overlapping rules are not evaluated top to bottom but non-deterministically. When we check compNat, Sloth reports the following counter-examples.

```
> strictCheck compNat 5
17: \(0 IHi) (I \perp) -> LT
20: \(I \perp) (0 IHi) -> GT
Finished 61 tests.
```

The first counter-example states that compNat yields $\perp$ if we apply it to $O I H i$ and $I \perp$, but there exists a less strict implementation that yields $L T$ instead. And, indeed, if we compare $O I H i$ with $I \perp$, that is, if we compare 2 with $2 * n+1$ where n is a natural number greater or equal to 1 , the former one is certainly always smaller then the latter one.

If we apply compNat to $O \mathrm{IHi}$ and $I \perp$, the function compNat compares IHi and $\perp$. In the case that the result of this test is $E Q$ as well as if it is $L T$ the result of compNat (OIHi) (I $\perp$ ) is LT. If the recursive application yields GT, then compNat yields GT as well. That is, instead of applying the function compNat recursively in the case that the arguments have the form ( $O \mathrm{~m}$ ) and ( $\mathrm{I} n$ ) we can check whether $m$ is less or equal to $n$. This way compNat makes use of the fact that the result of compNat is not $E Q$ if the outermost constructors already differ. Or in other words, if the outermost constructors differ we do not have to check whether the arguments of the constructors are equal. Using this idea we define the following less strict implementation of compNat. For readability we only present the rules that have changed in comparison with compNat.

```
compNat \({ }^{\prime}:\) Nat \(\rightarrow\) Nat \(\rightarrow\) Ordering
compNat \((\) O m) \((\operatorname{In})=\) if \(m<=n\) then \(L T\) else \(G T\)
compNat \({ }^{\prime}(\mathrm{Im}) \quad(O n)=\) if \(n<=m\) then \(G T\) else \(L T\)
```

Braßel et al. (2008) have defined $(<=)$ and $(>=)$ by means of compNat. In this case, compNat' is as strict as compNat. In the following we analyse the behavior of the comparison functions in more detail.

As Braßel et al. (2008) correctly observe, we "can do better [..] than defining

$$
\begin{aligned}
& n==m=i s E Q(\operatorname{compNat} n m) \\
& n /=m=\operatorname{not}(n==m)
\end{aligned}
$$

by sticking to the usual definition of $(==)$, which directly compares the constructors". In other words, if we implement the equality check ( $==$ ) by means of compNat (as well as compNat'), it is unnecessarily strict. For example, Sloth reports the following counter-examples if we check $(==)$.

```
> strictCheck (==) 5
10: \(\backslash(0 \perp)(\mathrm{I} \perp)=\) False
12: \(\backslash(\mathrm{I} \perp)(\mathrm{O} \perp)=\) False
17: \(\backslash(\mathrm{O}\) IHi) \((\mathrm{I} \perp)=\) False
20: \\(I IHi) \((0 \perp\) ) = False
Finished 61 tests.
```

The first counter-example is definitely a counter-example as for all total inputs, that are more defined than $O \perp$ and $I \perp$, the function $(==)$ yields False as well. Obviously, a naive implementation of $(==$ ) also yields False for the arguments $O \perp$ and $I \perp$. An implementation of $(==)$ that is based on compNat cannot show the same behavior as compNat has to evaluate one of its arguments completely to yield a result. For example, compNat cannot decide whether $O \perp$ is greater or less than $I \perp$.

Although Braßel et al. (2008) have observed that it is disadvantageous to define ( $==$ ) by means of compNat, they have not observed that the same holds for the other comparison functions. We define $(<),(>),(<=)$ and ( $>=$ ) by means of compNat like it is suggested by Braßel et al. (2008) and observe their behavior in the following. We consider the following definitions of the comparison functions as suggested by Braßel et al. (2008).

```
\(n<m=\) isLT (compNat \(n m\) )
\(n>m=\) isGT (compNat \(n m\) )
\(n<=m=\operatorname{not}(n>m)\)
\(n>=m=n o t(n<m)\)
```

In the following we only consider the function $(<=)$. We get similar results for the other functions. When we check $(<=)$ using Sloth, we get the following counter-example.

```
> strictCheck (<=) 4
2: \IHi \perp -> True
Finished 37 tests.
```

The natural number $I H i$ is less or equal to any other binary natural number. Therefore, we do not have to check the second argument. However, as $(<=)$ is defined by means of compNat it checks whether the second argument is IHi as in this case compNat yields $E Q$. If we increase the considered size, we observe that we also have $I \perp<=O I H i \equiv \perp$ while a minimally strict implementation yields False instead. As we have observed before, compNat is unnecessarily strict for the arguments $I \perp$ and $O I H i$ and, therefore, $(<=)$ is unnecessarily strict for these arguments as well. Thus, we have to provide a specific definition for $(<=)$ to get a minimally strict implementation.

In fact, this is a quite common problem that is not restricted to the comparison of binary numbers. For example, in Haskell the function $(<=)::$ Bool $\rightarrow$ Bool $\rightarrow$ Bool, which implements the Boolean implication, is unnecessarily strict. Sloth yields the following result.

```
> strictCheck ((<=) :: Bool -> Bool -> Bool) 3
2: \False }\perp=\mathrm{ True
Finished }7\mathrm{ tests.
```

In the same way as we expect False $\& \& \perp \equiv$ False we might expect False $<=\perp \equiv$ True. However, $(<=)$ is by default defined by means of compare. In other words, if we do not provide an implementation of $(<=)$ but an implementation of compare, then $(<=)$ is defined as follows.

$$
\begin{gathered}
(<=):: \text { Ord } \alpha \Rightarrow \alpha \rightarrow \alpha \rightarrow \text { Bool } \\
x<=y= \\
\text { case compare } x y \text { of } \\
\quad G T \rightarrow \text { False } \\
\quad-\quad \rightarrow \text { True }
\end{gathered}
$$

This implementation of $(<=)::$ Bool $\rightarrow$ Bool $\rightarrow$ Bool evaluates its second argument if its first argument is False.

In fact, we can show that every implementation of $(<=)$ that is implemented by means of compare for a data type that has a minimal element with respect to $(<=)$ is too strict. If the data type has a minimal element $m$ with respect to $(<=)$, we can implement $(<=)$ such that $m<=\perp \equiv$ True. In contrast, the Haskell function compare will always evaluate both arguments as they might be equal. Hence, if we implement $(<=)$ by means of compare we get $m<=\perp \equiv \perp$. Lastly, note that in most cases there is a minimal element with respect to $(<=)$. For example, all standard Haskell data types except for primitive data types like Int have a minimal element.

In Haskell, instead of only providing an implementation of compare we can as well only provide an implementation of $(<=)$. In this case compare is by default implemented by means of $(<=)$ as follows.

```
compare :: Ord \alpha=>\alpha->\alpha->Ordering
compare x }y=\mathrm{ if }x==y\mathrm{ then EQ else if }x<=y\mathrm{ then LT else GT
```

If so, both, $(<=)$ and compare are minimally strict, but compare potentially traverses its arguments twice because it first checks whether the arguments are equal and then checks whether they are related by $(<=)$. Therefore, it is advantageous to provide definitions for both functions, compare and $(<=)$. However, this is a very fine line because we have to trade modularity for a minimally strict implementation. Note that for all types whose $(<=)$ instance is unnecessarily strict the implementations of min and max are unnecessarily strict too as these functions are defined by means of ( $<=$ ) and the implementations of minimum and maximит are unnecessarily strict as these are defined by means of min and max.

Besides the Nat data, presented here, Braßel et al. (2008) present a data type for integer numbers that is based on Nat. A couple of functions for this integer type are not minimally strict because they are based on functions for Nat that are not minimally strict. Apart from these functions, subtraction of integers as well as division are not minimally strict.

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Figure 7.4.1: Heap Profile of Escaping Umlauts in Shakespeare's Collected Works using splitWhen from Version 0.1.2 of Split.

### 7.4 The split Package

In this section we check whether some functions of the Hackage package split, which provides a variety of functions to split lists into sublists, are minimally strict. We start by investigating the function splitWhen that splits a list by means of a predicate into a list of sublists.

First, we take a look at three different functions that solve this task, that is, we consider three functions of type $(\alpha \rightarrow$ Bool $) \rightarrow[\alpha] \rightarrow[[\alpha]]$, that behave equally with respect to total arguments.
splitWhen from version 0.1.4 of split by Yorgey (2011) In Section 3.2 we have used this implementation to define replaceBy. Using this function escaping an umlaut in Shakespeare's collected works uses constant space.
chop from the utility-ht package by Thielemann (2009) We have implemented the function replaceBy by means of chop when we have originally observed that intersperse is unnecessarily strict (Christiansen 2011). Using this function escaping an umlaut in Shakespeare's collected works uses constant space as well.
splitWhen from version 0.1.2 of split by Yorgey (2010) We did not use the function splitWhen when we originally observed that intersperse is unnecessarily strict, because splitWhen from version 0.1.2 of split, which has been the current version back then, causes a space leak itself. Using this function escaping an umlaut in Shakespeare's collected works does not use constant space at all.

Figure 7.4.1 presents the heap profile of escaping an umlaut in Shakespeare's collected works using splitWhen from version 0.1.2 of split. This task takes more than
ten minutes and uses around one hundred megabyte of memory. In contrast, using chop or the current version of splitWhen, this process takes less than one second and less than three kilobyte of memory. That is, this application takes around 600 times as long and more than 30,000 times the memory if we use the old implementation of splitWhen. Even in comparison to the results for the unnecessarily strict implementation of intersperse, the consequences of using the implementation of splitWhen from version 0.1.2 of split are worse.

By using Sloth we observed that some fundamental functions of the split package are unnecessarily strict. The improvements between version 0.1.2 and 0.1.4 of the package are due to improvements of functions with respect to strictness communicated to the author of the package by private communication. In Section 7.4.1, we show how we can observe that the original implementation of splitWhen (from version 0.1 .2 of split) is unnecessarily strict and improve its implementation. To improve splitWhen we improve a function that is the basis for most of the functions provided by the package split. That is, in fact, we do not only improve splitWhen but most of the functions provided by the package. In Section 7.4.2 we consider another function from split, called insertBlanks and improve it with respect to strictness.

Before we start, we want to state some preliminaries that are necessary to check a function using Sloth. The argument as well as the result type of the function have to be instances of the type classes Typeable and Data (Lämmel and Peyton Jones 2003). Haskell provides a mechanism to derive instances of these type classes by using the option DeriveDataTypeable. For example, using this option, to check the Peano multiplication from Section 7.2 we have to add the following line after the data type declaration for Peano.

## deriving (Typeable, Data)

In Section 3.2, when we checked intersperse, we did not have to add a deriving statement as lists as well as the data type $A$ are already instances of Typeable and Data.

Moreover, we can use so-called stand-alone deriving to generate these instances. More precisely, by using the option StandaloneDeriving we can derive instances of type classes without having to change the source code. That is, we can derive the required instances of data types that are defined in a library like split without changing the source code of the library. Without stand-alone deriving we would have to add the deriving statement to the data type declaration of the corresponding data type. The following code derives instances of Typeable and Data for the data type Chunk, which will be used in the following sections.
deriving instance Typeable1 Chunk
deriving instance Data $\alpha \Rightarrow$ Data (Chunk $\alpha$ )
Note that we have to derive an instance of Typeable1 instead of Typeable because Chunk is a type constructor. Furthermore, if a type $\tau$ is an instance of Typeable and a type constructor $\sigma$ is an instance of Typeable1, then $\sigma \tau$ is an instance of Typeable as well.

### 7.4.1 The Function splitWhen

After these preparations we would like to check if the function splitWhen is unnecessarily strict. Sadly, we cannot check whether this function is minimally strict as it has a higher-order type and Sloth does not support higher-order arguments. Therefore, we investigate the consequences of checking if splitWhen applied to a specific function is minimally strict. First, we check whether the partial application splitWhen (const False) is minimally strict. That is, we consider splitWhen applied to a predicate that is never satisfied. Sloth yields the following result.

```
> strictCheck (splitWhen (const False) :: [A] -> [[A]]) 3
1: \\perp -> \perp:[]
3: \(a:\perp) -> (a:\perp):[]
Finished 11 tests.
```

First of all both counter-examples stay counter-examples if we increase the considered size. The first counter-example states that splitWhen always yields a list with exactly one element. More precisely, if we apply splitWhen (const False) to $\perp$, we get $\perp$ while a minimally strict implementation yields $[\perp]$ instead. Obviously, splitWhen does not always yield a list with one element if we consider an arbitrary predicate. That is, the counter-example that shows that splitWhen (const False) is unnecessarily strict does not show that splitWhen is unnecessarily strict as well. Note that we cannot check which function splitWhen is applied to, and, therefore, we cannot define a function that satisfies the counter-examples if the functional argument of splitWhen is const False.

This example is supposed to demonstrate that we cannot check whether a high-er-order function is minimally strict by checking whether its partial application to an arbitrary function is minimally strict. Nevertheless, at least in the case of splitWhen there exists a predicate that allows us to derive some useful informations. We use the identity function as predicate. In this case the elements of the list determine the result of the predicate. Note that we can only use id as a predicate because splitWhen is polymorphic. If we consider a monomorphic function similar to splitWhen, for example, of type (Int $\rightarrow$ Bool) $\rightarrow[$ Int $] \rightarrow[[$ Int $]]$, we cannot use id as first argument. Instead we would have to define a function that interprets integers as Boolean values.

We get the following results if we check whether splitWhen id is minimally strict.

```
> strictCheck (splitWhen id) 4
1: \\perp -> \perp: \perp
3: \(\perp:\perp) -> \perp:\perp
4: \(False: }\perp\mathrm{ ) -> (False: }\perp\mathrm{ ) : }
5: \(True: \perp) -> []: \perp:\perp
Finished 13 tests.
```

First of all, all counter-examples stay counter-examples if we increase the size. The second question to answer is whether the counter-examples depend on the specific predicate again. Let us consider the counter-example labeled with four. By evaluating the first element of the list, we are always able to determine whether the first
element satisfies the predicate no matter which predicate we consider. Therefore, if we apply splitWhen $p$ to False $: \perp$ independent, of the predicate $p$ we know whether the first element satisfies the predicate. In the case of the predicate $i d$ the first element does not satisfy the predicate because the first element is False. In this case, obviously, any implementation of splitWhen is supposed to be able to yield the element for which the predicate is not satisfied. For example, consider the standard Haskell function words :: String $\rightarrow$ [String] that takes a string and splits this string at all occurrences of a whitespace. The following equivalence resembles the behavior that Sloth proposes for splitWhen.

$$
\text { words }\left('^{\prime}: \perp\right) \equiv\left(\prime^{\prime}: \perp\right): \perp
$$

That is, the presented counter-example is independent of the specific predicate.
To improve the implementation of splitWhen with respect to strictness we examine its implementation. The function splitWhen is defined as follows.

$$
\begin{aligned}
& \text { splitWhen }::(\alpha \rightarrow \text { Bool }) \rightarrow[\alpha] \rightarrow[[\alpha]] \\
& \text { splitWhen }=\text { split } \circ \text { dropDelims } \circ \text { whenElt }
\end{aligned}
$$

The function whenElt takes a predicate and yields a kind of configuration, and the function dropDelims enriches this configuration with additional information. The additional information is used to tell split that we want to drop the list elements that satisfy the predicate. As dropDelims and whenElt do not perform any pattern matching on the list, they cannot be the reason why splitWhen is unnecessarily strict. Therefore, we consider the implementation of split.

```
split:: Splitter \alpha }->[\alpha]->[[\alpha]
split s = map fromElem }\circ\mathrm{ postProcess s o splitInternal (delimiter s)
```

The argument of type Splitter $\alpha$ is the configuration that, among other things, contains the predicate that is used to split the list. Without going into details about the definition of split let use consider the implementation of splitInternal, which is the basis for most functions defined in the package split.

```
splitInternal:: Delimiter \alpha }->[\alpha]->[\mathrm{ Chunk }\alpha
splitInternal_[] = []
splitInternal d xxs@(x:xs)=
    case matchDelim d xxs of
        Just ([],(r:rs)) -> Delim []:Text [r]:splitInternal d rs
        Just (match, rest) }->\mathrm{ Delim match: splitInternal d rest
                            x'consText' splitInternal d xs
```

```
consText :: \alpha }->[\mathrm{ Chunk }\alpha]->[\mathrm{ Chunk }\alpha
```

consText :: \alpha }->[\mathrm{ Chunk }\alpha]->[\mathrm{ Chunk }\alpha
consText z (Text c:ys) = Text (z:c):ys
consText z (Text c:ys) = Text (z:c):ys
consTextzys}=\mathrm{ Text [z]:ys

```
consTextzys}=\mathrm{ Text [z]:ys
```

When we apply splitWhen to the predicate $i d$, the function splitInternal is applied to DelimEltPred id where DelimEltPred is a wrapper that takes a predicate and yields

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a value of the type Delimiter $\tau$. To examine why the application splitWhen id is not minimally strict, we check whether splitInternal (DelimEltPred id) is minimally strict.

When we check splitInternal (DelimEltPred id), we get similar results as we have got for splitWhen id. Note that later the postprocessing of the result of splitInternal will remove the Text constructors and drop the Delim elements. Sloth yields the following results.

```
> strictCheck (splitInternal (DelimEltPred id)) 5
3: \( }\perp:\perp) -> \perp:
4: \(False:\perp) -> Text (False: }\perp):
7: \(False: }\perp:\perp) -> Text (False: \):
9: \(True: }\perp:\perp) -> Delim (True: []):\perp:
Finished 21 tests.
```

As expected, the function splitInternal is not minimally strict. So, why is the application splitInternal (DelimEltPred id) (False : $\perp$ ) unnecessarily strict? The function splitInternal passes its arguments to matchDelim. By evaluating the application matchDelim (DelimEltPred id) (False : $\perp$ ) we observe that it yields Nothing. Thus, the first two cases of the pattern matching in splitInternal do not match but the third one does. More precisely, the considered application of splitInternal yields the result of consText False (splitInternal (DelimEltPred id) $\perp$ ). Furthermore, the following equivalence holds.

```
consText False (splitInternal (DelimEltPred id) \perp)
    { definition of splitInternal }
consText False }
    {definition of consText }
\perp
```

As the result of matchDelim is Nothing, and, therefore, cannot be less strict, we check whether consText is minimally strict. Note that we can use the more efficient type $A$ in this case as consText is polymorphic. Moreover, we do not have to bother about higher-order as consText is a first-order function. Sloth yields the following coun-ter-examples for consText.

```
> strictCheck (consText :: A -> [Chunk A] -> [Chunk A]) 3
1: \a \perp -> Text (a:\perp):\perp
3: \a ( }\perp:\perp) -> Text (a:\perp):
Finished 5 tests.
```

Sloth reports that consText is unnecessarily strict. The first element of the result list is always a Text element. Nevertheless, the implementation performs pattern matching on its argument before it yields the first result element.

By replacing pattern matching by a case expression we get the following equivalent implementation of consText. This implementation shows that we can apply case deferment to derive a less strict implementation.

```
consText:: \alpha -> [Chunk \alpha] ->[Chunk \alpha]
consText x cs =
```

```
case cs of
    Text text : rest }->\mathrm{ Text (x : text) : rest
    - }->\mathrm{ Text (x:[]):cs
```

However, in this case we would need a context with multiple holes, more precisely, the context $C=\operatorname{Text}(x:[\cdot]):[\cdot]$. Instead of generalizing case deferment to contexts with multiple holes we introduce a tuple that combines the contents of the holes to a single value. More precisely, we introduce let expressions on all right-hand sides of the case expression and get the following equivalent implementation of consText. In a similar way we can handle all cases of contexts with multiple holes by introducing a let expression.

```
consText \(:: \alpha \rightarrow[\) Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
consText \(x\) cs \(=\)
    case \(c s\) of
        Text text : rest \(\rightarrow \mathbf{l e t}\left(\right.\) text \(^{\prime}\), rest \(\left.{ }^{\prime}\right)=(\) text, rest \()\) in Text \((x:\) text \():\) rest \(^{\prime}\)
        _ \(\quad \rightarrow\) let \(\left(\right.\) text \(^{\prime}\), rest \(\left.^{\prime}\right)=([], c s) \quad\) in Text \(\left(x:\right.\) text \(\left.^{\prime}\right):\) rest \(^{\prime}\)
```

Now, we can use a context with a single hole, namely,

$$
C=\text { let }\left(\text { text } t^{\prime}, \text { rest } t^{\prime}\right)=[\cdot] \text { in Text }(x: \text { text }): \text { rest } .
$$

Furthermore, we use $D=[\cdot]$ and apply Lemma 7.1.1 to derive the following implementation by moving the context $C$ to the front of the definition.

$$
\begin{aligned}
& \text { consText }:: \alpha \rightarrow[\text { Chunk } \alpha] \rightarrow[\text { Chunk } \alpha] \\
& \text { consText } x^{\prime} \text { cs }= \\
& \text { let }\left(\text { text }^{\prime}, \text { rest }{ }^{\prime}\right)=\text { case } c s \text { of } \\
& \text { Text text : rest } \rightarrow \text { (text, rest) } \\
& -\quad \rightarrow([], c s) \\
& \text { in Text ( } x: \text { text }^{\prime} \text { ) : rest }{ }^{\prime}
\end{aligned}
$$

To show that this implementation is less strict we consider the arguments $x=a$ where $a$ is an arbitrary value of some type $\tau$ and $c s=\perp$. In this case, we have $c s \equiv \perp$ and

```
D[C[\perp]]
    \equiv{definition of C and D }
let (text',rest')=\perp in Text (a:text'):rest'
    \equiv{ semantics of let expression }
Text (a:\perp):\perp
    \square
```

Thus, consText ${ }^{\prime}$ is less strict than consText. Note that we do not have to consider the definition of consText by means of an explicit fixpoint to apply case deferment as its definition is non-recursive.

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Figure 7.4.2: Heap Profile of Escaping Umlauts with the Less Strict Implementation of splitWhen.

As we will reuse it later, we define a function splitText at top-level instead of using a case expression and use the following definition of consText ${ }^{\prime}$ that is equivalent to the derived implementation.

```
consText \(:: \alpha \rightarrow[\) Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
consText' \(x\) cs \(=\) Text \((x\) : text \():\) rest
    where
        \((\) text, rest \()=\) splitText cs
splitText :: [Chunk \(\alpha] \rightarrow([\alpha],[\) Chunk \(\alpha])\)
splitText \((\) Text text \(:\) rest \()=(\) text,rest \()\)
splitText cs \(\quad=([], c s)\)
```

Figure 7.4.2 shows the heap profile of escaping umlauts if we use this less strict implementation of consText and, therefore, a less strict implementation of splitWhen. This profile is quite similar to the profile that we get for chop (Thielemann 2009) and splitWhen from version 0.1.4 of the split package (Yorgey 2011).

So, why do we get a significant performance improvement by improving the inconspicuous looking function consText? When we take a closer look at the coun-ter-example, we observe that if the predicate yields False for all list elements, then splitInternal does not yield a first Text element until the list is terminated by the empty list. For example, for the function splitInternal from version 0.1 .2 of split we have

$$
\text { splitInternal (DelimEltPred id) (replicate n False }+\perp \text { ) } \equiv \perp
$$

for all $n$ :: Int. That is, we have to completely evaluate the list structure to get the first element of the result list. As we consider a predicate that is not satisfied for any list element - we check whether any character of Shakespeare's collected works is the
umlaut "ä" -, we have to completely evaluate the argument of splitWhen, which is the string containing Shakespeare's collected works. If we take another look at the profile in Figure 7.4.1, we observe that the heap is filled by TSO elements. TSO is the abbreviation of Thread State Object. Although the name includes "Thread", these objects also occur in non-threaded programs where they refer to the stack usage of the main thread. In this particular example, by being unnecessarily strict, the compiler has to evaluate a cascade of non-tail recursive applications and to allocate stack space for all these applications. If the first element of a list does not satisfy the predicate, then an application of splitInternal reduces to an application of consText to the first element of the list and a recursive application of splitInternal to the rest of the list. However, consText does not yield a result before it has evaluated its second argument, namely, the application of splitInternal to the rest of the list. Thus, if all elements of a list do not satisfy a predicate, we have to allocate stack space for as many applications of splitInternal as the list has elements. In the case of Shakespeare's collected works the compiler allocates stack space for as many applications of splitInternal as Shakespeare has written down characters.

As we have observed, the function splitWhen is another example of a function whose performance can be improved by improving its implementation with respect to strictness. Yet, there are many ways to interfere with the memory behavior even if a function is minimally strict. For example, an implementation of splitWhen that is based on consText ${ }^{\prime}$ causes a space leak if a compiler would use a naive approach to the implementation of lazy pattern matchings like the where clause in consText ${ }^{\prime}$. Wadler (1987) has presented an optimization that prevents this kind of space leak.

To get the basic idea of this optimization let us consider the naive approach using the example of consText ${ }^{\prime}$. The naive approach introduces a variable, called pair, for the result of the application of splitText. Furthermore, it replaces the variable text by fst pair and rest by snd pair, where fst and snd are projections to the first and the second component of a pair, respectively. That is, we get the following implementation.

$$
\begin{aligned}
& \text { consText }{ }_{2}^{\prime}:: \alpha \rightarrow[\text { Chunk } \alpha] \rightarrow[\text { Chunk } \alpha] \\
& \text { consText } x \text { cs }=\text { Text }(x: \text { fst pair }): \text { snd pair } \\
& \text { where } \\
& \quad \text { pair }=\text { splitText cs }
\end{aligned}
$$

To cause a space leak we have to consider a context that first evaluates the first list element of the result of consText ${ }_{2}^{\prime}$ and afterwards evaluates the rest of the list. When the application $f_{s t}$ pair is evaluated, the application splitText cs is evaluated. Although, the function does not reference the first component of the pair anymore, the garbage collector is not able to deallocate the corresponding memory as the program still keeps a reference to the whole pair via the application snd pair. To overcome this problem, intuitively, when fst pair is evaluated occurrences of snd pair are replaced by a shortcut to the second component of the tuple. This way, there is no reference that keeps the whole pair alive. The interested reader may consider the work by Wadler (1987) and Sparud (1993) for more details about this problem and its solution.

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Although this optimization prevents consText from causing a space leak, we can use a denotationally equivalent implementation of consText that does not allow for this kind of optimization. For example, consider the following function consText ${ }_{3}^{\prime}$.

```
consText \({ }_{3}^{\prime}:: \alpha \rightarrow[\) Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
consText \({ }_{3}^{\prime} x\) cs \(=\) Text ( \(x\) : text cs) : rest cs
text (Text \(\left.t:{ }_{2}\right)=t\)
text \(\quad=[]\)
rest \((\) Text _: cs \()=c s\)
rest cs \(=c s\)
```

If we define splitWhen by means of consText ${ }_{3}^{\prime}$, escaping an umlaut uses less memory than using the original definition of splitWhen, but it uses significantly more memory than using an implementation of splitWhen that is based on consText ${ }^{\prime}$.

Let us consider a context as we have done above, that is, a context that first evaluates the first list element of the result of consText ${ }_{3}^{\prime}$ and afterwards evaluates the rest of the list. If the first list element of the result is evaluated, the application text cs is evaluated. Although the function text projects to a part of the argument of cs that is not used later on, the garbage collector is not able to deallocate the occupied memory because the program still has a reference to the argument list $c s$. Instead, the function holds the part of cs that has been evaluated until the expression rest cs is evaluated.

This example reveals a general problem of minimally strict functions. From the point of the less-strict relation, as defined in this thesis, the functions consText ${ }^{\prime}$ and consText ${ }_{3}^{\prime}$ are equivalent as we use a denotational view on functions. That is, among all minimally strict implementations we want to chose the implementation that uses the smallest amount of memory. Obviously, an operational view on functions is desirable if we want to reason about memory behavior. In Section 7.5 we will observe that even with a denotational model we can still draw some connections between the less-strict relation and memory usage.

Let us switch back to the undertaking of defining a minimally strict implementation of splitWhen, we observe that, in fact, splitInternal and, thus, splitWhen are still unnecessarily strict. By splitInternal' we denote an implementation of splitInternal that uses the minimally strict implementation consText instead of the original implementation consText. Sloth reports the following counter-examples for splitinternal'.

```
> strictCheck (splitInternal' (DelimEltPred id)) 2
3: \( }\perp:\perp) -> \perp:
Finished 5 tests.
```

If we take a second look at the definition of splitInternal, we observe that all righthand sides of the case expression yield a list with at least one element. In other words, there is a common context that occurs on all right-hand sides of the case expression. Hence, we can use case deferment to derive the following less strict definition.

```
splitInternal"
splitInternal" _ [] = []
```

```
splitInternal" \(d x x s @(x: x s)=h d: t l\)
    where
        \((h d, t l)=\) case matchDelim \(d x x s\) of
                            Just \(([],(r: r s)) \rightarrow(\) Delim [], Text [r]:splitInternal" d rs)
                            Just (match, \(r s) \rightarrow\) (Delim match, splitInternal" \(d r s)\)
                            \(\rightarrow\) (Text \((x\) : text \()\),rest \()\)
        \((\) text, rest \()=\) splitText \(\left(\right.\) splitInternal \(\left.{ }^{\prime \prime} d x s\right)\)
```

However, this implementation is still unnecessarily strict as the following coun-ter-example demonstrates.

```
> strictCheck (splitInternal', (DelimEltPred id)) 3
6: \(\perp:[]) -> \perp:[]
Finished 13 tests.
```

At first sight, this counter-example might look as it contradicts the sequentiality of splitInternal ${ }^{\prime \prime}$ as we have splitInternal" (False $\left.: \perp\right) \equiv$ Text (False $\left.: \perp\right): \perp$. That is, we are supposed to perform pattern matching on the head of the list and the rest of the list. The counter-example does not contradict sequentiality as we are supposed to perform pattern matching on the head and the tail of the list with respect to different result positions. With respect to the head of the result list, we perform pattern matching on the head of the argument list and with respect to the tail of the result list, we perform pattern matching on the tail of the argument list.

The application splitInternal ${ }^{\prime \prime}$ (DelimEltPred id) ( $\perp$ :[]) applies matchDelim to the arguments DelimEltPred id and $\perp$ : []. This application yields $\perp$ and the result of splitInternal" is $\perp: \perp$. To define a less strict implementation, we add an additional check whether the tail of the argument list is the empty list. We replace the expression

## head: tail

by

$$
\text { head : if null } x \text { s then [] else tail. }
$$

Note that this new implementation is not more strict than the original one. If we have $x s \equiv \perp$, then tail $\equiv \perp$, and, therefore,

$$
\text { if null } x s \text { then [] else tail } \equiv \perp \equiv \text { tail. }
$$

However, if $x \equiv \perp$ and $x s \equiv[]$, then tail $\equiv \perp$ and, thus,

$$
\text { if null } x s \text { then }[] \text { else tail } \equiv[] \sqsupset \text { tail. }
$$

### 7.4.2 The Function insertBlanks

There are several other functions in the split package that are unnecessarily strict. In the following, we exemplarily consider the function insertBlanks. It takes a list of Chunks, that is, a list of Text and Delim elements and inserts blanks, namely, Text elements with an empty list, between all pairs of succeeding delimiters. Furthermore,

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the function adds a blank at the end of the list, and if the first list element is a delimiter the function adds a blank at the front. We do not discuss the implementation of insertBlanks in detail as we will only apply purely syntactical transformations in the following.

```
insertBlanks :: \([\) Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
insertBlanks [] \(=[\) Text []]
insertBlanks \((d @(\) Delim _) \(): l)=\) Text []:insertBlanksRest \((d: l)\)
insertBlanks \(l=\) insertBlanksRest \(l\)
insertBlanksRest :: [Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
insertBlanksRest [] = []
insertBlanksRest \([\) d@(Delim \()] \quad=[d\), Text []]
insertBlanksRest (d1@(Delim _) :d2@(Delim _) :l)=
    d1:Text []: insertBlanksRest (d2:l)
insertBlanksRest \((c: l)=c:\) insertBlanksRest \(l\)
```

When we check insertBlanks using Sloth, we get the following result.

```
> strictCheck (insertBlanks :: [Chunk A] -> [Chunk A]) 2
1: \_ -> Text _:_
3: \(_:_) -> Text _:_
Finished 5 tests.
```

The first counter-example states that the first element of the result of insertBlanks is always a Text element. Nevertheless, insertBlanks first evaluates its argument before it yields the first element of the result list.

We start with a simple transformation of the definition of insertBlanks. The data type Chunk provides two constructors, namely, Text and Delim, which both take a list as argument. Hence, in the last rule of insertBlanks the variable $l$ has the form Text text : l. Besides this observation, we use the following equality

$$
\text { insertBlanksRest }(\text { Text text }: l) \equiv \text { Text text }: \text { insertBlanksRest } l
$$

and arrive at an equivalent definition of insertBlanks.

```
insertBlanks :: [Chunk \alpha] -> [Chunk \alpha]
insertBlanks[] =[Text []]
insertBlanks (d@(Delim _) :l) = Text []:insertBlanksRest (d:l)
insertBlanks (Text text:l) = Text text:insertBlanksRest l
```

To derive a less strict implementation of insertBlanks we consider the following equivalent implementation of insertBlanks that uses a case expression. Furthermore, we use the equivalence insertBlanksRest []$\equiv[]$ and change the order of the rules in the case expression.

```
insertBlanks:: [Chunk \alpha] -> [Chunk \alpha]
```

insertBlanks cs =
case $c s$ of
Text text : $l \rightarrow$ Text text : insertBlanksRest $l$
$l \rightarrow$ Text [] :insertBlanksRest $l$
Now, we can apply case deferment to derive a less strict implementation. We consider the contexts $D=[\cdot]$ and $C=$ Text $[\cdot]:[\cdot]$ and apply Lemma 7.1.1. As we consider a context with multiple holes, we have to introduce a let expression, but we abstain from developing the implementation step by step and simply present the final result. After applying case deferment we observe that we can reuse splitText and get the following less strict implementation.

```
insertBlanks' \(::[\) Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
insertBlanks' \(l=\) Text text \(:\) insertBlanksRest rest
    where
        \((\) text, rest \()=\) splitText \(l\)
```

However, this new implementation of insertBlanks is still unnecessarily strict as the following result shows.

```
> strictCheck (insertBlanks' :: [Chunk A] -> [Chunk A]) 4
8: \(Delim _:_) -> Text []:Delim _:Text _:_
13: \(Delim _:_:_) -> Text []:Delim _:Text _:_
Finished 23 tests.
```

Because the counter-examples only affect the tail of the result list, probably the function insertBlanksRest is too strict. We verify this assumption by checking whether insertBlanksRest is minimally strict.

```
> strictCheck (insertBlanksRest :: [Chunk A] -> [Chunk A]) 4
3: \(_:_) -> _:_
4: \(Delim _:_) -> Delim _:Text _:_
7: \(Delim _:_:_) -> Delim _:Text _:_
Finished }19\mathrm{ tests.
```

And, indeed, we get similar results when we check insertBlanksRest. Thus, we examine insertBlanksRest in detail. As the definition of insertBlanksRest is quite unwieldy, we define the following equivalent implementation of insertBlanks by introducing a case expression to perform pattern matching if the first list element is a delimiter. We replace the second and third rule of insertBlanksRest by the following rule.

```
insertBlanksRest (d1@(Delim _) :l1) =
    \(d 1\) : case 11 of
            [] \(\rightarrow\) [Text []]
            \(d 2 @(\) Delim _) \(: l \rightarrow\) Text []:insertBlanksRest (d2:l)
            \(c: l \quad \rightarrow c:\) insertBlanksRest \(l\)
```

When we take a closer look at the case expression of this rule, we observe that it is equivalent to the original definition of insertBlanks. Thus, we can replace the case expression by an application of insertBlanks' as follows.

```
insertBlanksRest :: [Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
insertBlanksRest' []\(=[]\)
insertBlanksRest \((d @(\) Delim _ \(): l)=d:\) insertBlanks' \(l\)
insertBlanksRest \((c: l) \quad=c:\) insertBlanksRest' \(l^{l}\)
```

Now, we are able to apply case deferment and get the following less strict implementation. For simplicity, we replace the case expression by a predicate that checks whether its argument is a delimiter, called isDelim.

```
insertBlanksRest \(::[\) Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
insertBlanksRest' []\(=[]\)
insertBlanksRest' \((c: l)=\)
    \(c:\) if isDelim \(c\) then insertBlanks \({ }^{\prime} l\) else insertBlanksRest' \(l\)
```

We have presented insertBlanks for a particular reason. The memory performance of replacing umlauts in Shakespeare's collected works using the less strict implementation of insertBlanks is worse than the memory behavior if we use the original implementation. As we have mentioned before, the compiler applies an optimization to prevent a space leak in the presence of lazy pattern matching as it is used in the definition of insertBlanks ${ }^{\prime}$. However, in some cases, the current implementation of this optimization implemented in the GHC fails (Marlow 2008). More precisely, by applying other optimizations first, the compiler is unable to observe that the resulting program projects to the components of a tuple (Felgenhauer 2010). We can always prevent the space leak, caused by a lazy pattern matching, by replacing it by a case expression as follows.

```
insertBlanks \({ }_{2}^{\prime}::[\) Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
insertBlanks \({ }_{2}^{\prime} l=\)
    case splitText \(l\) of
        (text,rest) \(\rightarrow\) Text text: insertBlanksRest' rest
```

However, this implementation is too strict again, as we have insertBlanks ${ }_{2}^{\prime} \perp \equiv \perp$. Yet, there even exists an implementation that does not have a space leak and is still minimally strict. We use a mixture of a lazy pattern matching by means of a where clause and a strict pattern matching on the tuple. In the original definition, the compiler is unable to apply the optimization because we apply a function to one of the components of the tuple. Therefore, we introduce a case expression that applies this function to one of the components of the tuple but yields the complete tuple structure. The resulting tuple is destructed lazily by a where clause. More precisely, we consider the following function.

```
insertBlanks \({ }_{3}^{\prime}::[\) Chunk \(\alpha] \rightarrow[\) Chunk \(\alpha]\)
insertBlanks \({ }_{3}^{\prime} l=\) Text text \(t^{\prime}\) rest
    where
        (text \({ }^{\prime}\), rest \(\left.^{\prime}\right)=\) case splitText \(l\) of
            (text, rest) \(\rightarrow\) (text, insertBlanksRest' rest)
```

Because we do not apply a function to the variable rest' anymore, the compiler is able to identify the tuple selectors in contrast to the original minimally strict definition insertBlanks'. We did not investigate whether this transformation can be applied in general when using the current version of the GHC to derive an implementation that does not cause the space leak characterized by Wadler (1987).

### 7.5 Reversing Lists

When we considered memory improvements that arise from less strict implementations in the previous sections, we have always considered only one specific application. We have demonstrated that minimally strict implementations of intersperse and split When are advantageous with respect to memory usage when we consider a program that replaces all occurrences of an umlaut in Shakespeare's collected works. In this section we discuss the memory behavior of two minimally strict implementations in comparison to the original implementation with respect to different contexts. We consider three different implementations of reverse and examine their behavior with respect to three contexts. We will observe that the space usage of an implementation depends highly on the context it is used in. Gustavsson and Sands $(1999,2001)$ have provided a formal model for reasoning about space usage in a lazy language, but we will only informally argue about the space usage of a function here.

To begin with, consider the following linear complexity implementation of reverse.

```
reverse :: \([\alpha] \rightarrow[\alpha]\)
reverse \(=\) rev []
    where
        revexs [] \(=x s\)
        rev \(x s(y: y s)=\operatorname{rev}(y: x s) y s\)
```

When we check reverse using Sloth we, somehow surprisingly, observe that it is not minimally strict.

```
> strictCheck (reverse :: [A] -> [A]) 4
3: \(a:\perp) -> }\perp:
5: \(a:b:\perp) -> \perp: }\perp:
Finished 9 tests.
```

These counter-examples state that, although the argument list is terminated by $\perp$, reverse is supposed to yield the spine of the result list. If we apply reverse to a total list that is more defined than $a: \perp$, that is, to a list with at least one element, the result is always a list with at least one element. Therefore, the potential counter-examples are indeed counter-examples. However, every function that uses an accumulator parameter, like reverse, yields $\perp$ if the argument list is terminated by $\perp$.

We can define a minimally strict implementation of reverse as follows. The function last $::[\alpha] \rightarrow \alpha$ yields the last element of a non-empty list and the function init $::[\alpha] \rightarrow[\alpha]$ yields a list without this last element.

```
reverse \(::[\alpha] \rightarrow[\alpha]\)
reverse \({ }^{\prime}[]=[]\)
reverse \(x s=\) last \(x \mathrm{~s}:\) reverse \({ }^{\prime}(\) init \(x s)\)
```

This implementation yields $\perp: \perp$ for the argument $a: \perp$ as the following equivalence shows.

```
reverse' (a:\perp)
{ {definition of reverse' }
last (a:\perp): reverse' (init (a:\perp))
    \equiv{definition of last }
\perp:\mp@subsup{reverse}{}{\prime}(\mathrm{ init ( a: }\perp))
\equiv{definition of init, namely rule init [x]=[]}
|:reverse 
{definition of reverse }\mp@subsup{}{}{\prime}
\perp:\perp
```

While reverse' is minimally strict, in contrast to reverse it has a quadratic complexity. Note that the naive quadratic complexity implementation of reverse by means of ( + ) is not minimally strict.

Though, there is even a linear complexity implementation of reverse that is minimally strict. The result of reverse has as many list constructors as its argument. We define a function withShapeOf, that takes the result of reverse and its argument and reconstructs the spine of the result by performing pattern matching on the argument. By using a lazy pattern matching, the function withShapeOf only performs pattern matching on $l$ if any of the elements of the result of withShapeOf is evaluated.

```
withShapeOf \(::[\alpha] \rightarrow[\beta] \rightarrow[\alpha]\)
_'withShapeOf'[] \(=[]\)
\(l\) 'withShapeOf' (-:ys) \(=x:\left(x s^{\prime}\right.\) withShapeOf' \(\left.y s\right)\)
        where \(x: x s=l\)
reverse \({ }^{\prime \prime}::[\alpha] \rightarrow[\alpha]\)
reverse \({ }^{\prime \prime}\) xs \(=\) reverse \(x s\) 'withShapeOf'xs
```

The function reverse ${ }^{\prime \prime}$ is also minimally strict as the following equivalence illustrates.

```
reverse \({ }^{\prime \prime}(a: \perp)\)
    \(\equiv\left\{\right.\) definition of reverse \(\left.{ }^{\prime \prime}\right\}\)
reverse \((a: \perp)\) 'withShapeOf' \((a: \perp)\)
    \(\equiv\{\) definition of withShapeOf \(\}\)
head (reverse \((a: \perp)):(\) tail (reverse \((a: \perp))\) 'withShapeOf‘ \(\perp\) )
\(\equiv\{\) definition of reverse \(\}\)
head (reverse \((a: \perp)):(\) tail \(\perp\) 'withShapeOf' \(\perp)\)
    \(\equiv\) \{ definition of tail \}
head (reverse \((a: \perp)):(\perp\) 'withShapeOf' \(\perp)\)
    \(\equiv\{\) definition of withShapeOf \(\}\)
```

```
head (reverse \((a: \perp)): \perp\)
    \(\equiv \quad\{\) definition of reverse \(\}\)
head \(\perp: \perp\)
    \(\equiv\{\) definition of head \(\}\)
\(\perp: \perp\)
```

Although the function withShapeOf might look somehow artificial, Chitil (2005) uses an equivalent function, called copyListStructure, to reimplement an imperative pretty printing algorithm in a functional language. By cleverly employing the function copyListStructure, the resulting algorithm has the same complexity as the imperative algorithm (Wadler 2003).

Finally, note that there are other functions that show a similar behavior as reverse. For example, a function that takes a binary tree and yields the list of elements in the tree in in-order shows a similar behavior. A minimally strict implementation of this function is supposed to yield the first list constructor even if it is applied to a tree whose left children is undefined.

As we have observed in the previous sections, there are minimally strict implementations that use a non-constant factor less space than their unnecessarily strict counterparts with respect to a specific context as it is the case for intersperse and splitWhen. Here, we will argue that there always exists a context such that a less strict implementation uses a non-constant factor less space than the unnecessarily strict implementation.

So, which context might be a candidate for a non-constant improvement? The counter-example that shows that a function is unnecessarily strict provides a context in which the less strict implementation uses less space than the original implementation. For example, reverse is not minimally strict while reverse' is minimally strict because we have

$$
\text { reverse }(\perp: \perp) \equiv \perp
$$

but

$$
\text { reverse }^{\prime}(\perp: \perp) \equiv \perp: \perp .
$$

As context we consider a function, that yields $\perp$ if we apply it to the result of the unnecessarily strict implementation. In contrast, the function yields the empty tuple, denoted by (), if we apply it to the result of the less strict implementation. For example, in the case of reverse we consider the following function context.

$$
\begin{aligned}
& \text { context }::[\alpha] \rightarrow() \\
& \operatorname{context}(-:-)=()
\end{aligned}
$$

Then we have

$$
\text { context }(\text { reverse }(\perp: \perp)) \equiv \perp
$$

but

$$
\text { context }\left(\text { reverse }^{\prime}(\perp: \perp)\right) \equiv()
$$

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Now we look for an argument that is more defined than $\perp: \perp$ such that both applications are evaluated to (). For example, we have

$$
\text { context }(\text { reverse }(\perp:[])) \equiv()
$$

and

$$
\text { context }\left(\text { reverse }{ }^{\prime}(\perp:[])\right) \equiv() .
$$

This way we observe that, in comparison to reverse ${ }^{\prime}$, the function reverse inspects a larger part of its argument to yield the same result. More precisely, reverse evaluates the second argument of the cons constructor, in this case the empty list, to yield the first list constructor as result. Note that in Chapter 1 we characterized the less-strict relation exactly this way, that is, by referring to the inspected part of an argument.

Next, we define a function that yields the first argument, namely, $\perp: \perp$, in constant space while it needs linear space in the size of some integer to yield the more defined argument, in this case $\perp$ : []. For example, we can use the following definition, which is inspired by the motivating example by Gustavsson and Sands (1999).

```
argument :: Int \(\rightarrow[\alpha]\)
argument \(n=\perp\) : if \(y==x\) then \(\perp\) else []
    where
        \(x s=[0 \ldots n]\)
        \(x=\) head \(x s\)
        \(y=\) last \(x\) s
```

We assume that $n$ is always greater or equal to one. Furthermore, in this context by $\perp$ we always denote a run-time error and not a non-terminating expression as the evaluation of a non-terminating expression might result in space usage while this is not the case for a run-time error. Finally, the following considerations are based on the fact that the equality check $(==)$ on integers first evaluates its first argument.

To evaluate the expression if $y==x$ then $\perp$ else [] we first have to evaluate $y$ due to the left to right evaluation order of $(==)$. This causes the evaluation of the whole list structure of $x s$ and, therefore, occupies linear space in the size of $n$. The garbage collector cannot deallocate the occupied memory as we hold a reference to $x s$ by means of $x$.

Now we consider the following application.

$$
\text { context (reverse } \left.{ }^{\prime}(\text { argument } n)\right)
$$

The evaluation of this application has a constant space usage as reverse ${ }^{\prime}$ does not evaluate the conditional. In contrast, the evaluation of
context (reverse (argument n))
causes the evaluation of the expression if $y==x$ then $\perp$ else []. The evaluation of this expression uses a linear amount of space in the size of $n$. Therefore, by considering an arbitrary large integer $n$, we get a context in which reverse uses an arbitrary

| context | reverse | reverse $^{\prime}$ | reverse $^{\prime \prime}$ |
| :---: | :---: | :---: | :---: |
| null | $1.90 \mathrm{~s} / 460 \mathrm{MB}$ | $0.001 \mathrm{~s} / 44 \mathrm{~KB}$ | $0.001 \mathrm{~s} / 44 \mathrm{~KB}$ |
| head | $2.65 \mathrm{~s} / 460 \mathrm{MB}$ | $0.41 \mathrm{~s} / 28 \mathrm{~KB}$ | $2.65 \mathrm{~s} / 460 \mathrm{MB}$ |
| double | $2.99 \mathrm{~s} / 460 \mathrm{MB}$ | $>1 \mathrm{~min} />2 \mathrm{~GB}$ | $10.29 \mathrm{~s} / 2 \mathrm{~GB}$ |

Figure 7.5.1: Run-Times and Space Usage for the Evaluation of an Application of a Reverse Function to [ $1 . .2 * 10^{\wedge} 7$ ] in the Corresponding Context.
large amount of space while reverse' uses only a constant amount of space. The same holds if we use reverse ${ }^{\prime \prime}$ instead of reverse'.

In the same way we can derive a context in which an arbitrary less strict function is an improvement over its unnecessarily strict counterpart. For example, for intersperse we can use a quite similar construction as for reverse. More precisely, the evaluation of the application

$$
\text { context }(\text { intersperse } \perp(\text { argument } n))
$$

uses linear space while the evaluation of

$$
\text { context }\left(\text { intersperse }^{\prime} \perp(\text { argument } n)\right)
$$

uses only constant space. Obviously, as future work we have to formally verify these statements by using appropriate semantic models.

Instead of the artificial argument argument $n$ we can also use a more natural one, namely, $[1 \ldots n]$. That is, we do not have to attribute the linear amount of space to the evaluation of the empty list, but we can as well attribute it to the evaluation of a linear-sized list structure. Note that there are examples where we indeed have to attribute a linear amount of space to the evaluation of a single constructor. Thus, we have used the function argument above to illustrate that it is possible to attribute an arbitrary amount of space usage to the evaluation of a single constructor. Furthermore, instead of the context context, we can as well use the function null that checks whether its argument is the empty list. Figure 7.5 .1 presents run-times and space usages if we apply null to applications of reverse, reverse ${ }^{\prime}$, and reverse ${ }^{\prime \prime}$ to the list $\left[1 . .2 * 10^{\wedge} 7\right]$. Here, the term $\left[1 . .2 * 10^{\wedge} 7\right]$ denotes the list with the integers from one to 20 million. The numbers in Figure 7.5.1 together with results for smaller lists not presented here confirm that the unnecessarily strict implementation uses a linear amount of space in the length of the list, while the less strict implementations reverse ${ }^{\prime}$ and reverse ${ }^{\prime \prime}$ use only a constant amount of space.

Now let us consider another context, namely, head. As Figure 7.5.1 shows, the space usage of reverse' is quite good if we consider the context head, because the application head (reverse' $x s$ ) reduces to last $x s$ and last $[1 \ldots n]$ runs in constant space for all $n$. In contrast, the application head (reverse xs) constructs the reversed list completely. Therefore, reverse ${ }^{\prime}$ also uses a non-constant factor less space than reverse if we consider the context head. In contrast to reverse ${ }^{\prime}$, the application head (reverse ${ }^{\prime \prime}$ xs) reduces to head (reverse $x s$ ) and, thus, the space usage of reverse ${ }^{\prime \prime}[1 \ldots n]$ is linear in $n$ as

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it is the case for reverse. In the case of the context head the difference between reverse ${ }^{\prime}$ and reverse ${ }^{\prime \prime}$ with respect to space usage is not reflected by the less-strict ordering because reverse ${ }^{\prime}$ and reverse ${ }^{\prime \prime}$ are denotationally equivalent.

However, there even exists a context in which the space usage of reverse as well as the space usage of reverse ${ }^{\prime \prime}$ is quadratic in the size of the argument list while the space usage of reverse is linear in the size of the argument. We consider the function double. Its definition uses the predefined length function, which uses a strict evaluation primitive to evaluate intermediate results. If we do not evaluate intermediate results, length itself causes a space leak.

$$
\begin{aligned}
& \text { double }::[\alpha] \rightarrow \text { Int } \\
& \text { double ys }=\text { length }(y s+y s)
\end{aligned}
$$

When we evaluate the expression double (reverse ${ }^{\prime} x s$ ), the spine of reverse ${ }^{\prime} x s$ is evaluated by the application of length. Furthermore, length does not evaluate the elements of the list. Therefore, $y s$ is evaluated to a structure, that can be illustrated by the following expression.
last xs : last (init xs) : last (init (init xs)) : ...

The applications of last and init to the list $x$ s are unevaluated thunks. The garbage collector cannot collect this structure because in double we keep a reference to the list $y s$ in the second argument of ( + ). As this example illustrates, we get a quadratic number of thunks in the length of the list. According to the space model by Gustavsson and Sands (1999), a quadratic number of thunks occupies a quadratic amount of space. Thus, the evaluation of the application double (reverse ${ }^{\prime}[1 . . n]$ ) uses a quadratic amount of space in $n$.

The evaluation of the application double (reverse ${ }^{\prime \prime}[1 \ldots n]$ ) also uses a quadratic amount of space. In the case of reverse ${ }^{\prime \prime}$, unevaluated applications of last and init become unevaluated projections to the head and the tail of the list reverse $x \mathrm{~s}$. As these projections are never evaluated, we cannot benefit from the optimization proposed by (Wadler 1987) in this case.

This example demonstrates that the theory and practice of minimally strict functions lacks a more thorough investigation of space usage of minimally strict functions. Furthermore, it might be advantageous to aim for implementations that are minimally strict with respect to a specific context and not minimally strict with respect to all possible contexts. For example, if we check the function double o reverse :: $[A] \rightarrow$ Int, Sloth does not report any counter-examples. That is, if we use the context double we do not have to alter reverse because it is already minimally strict with respect to this context. However, if we consider the context null, that is, if we check whether null $\circ$ reverse $::[A] \rightarrow$ Bool is minimally strict, we observe that it is not. Nevertheless, checking whether the result of reverse is the empty list is an quite unusual application as we can perform the same check on the argument of reverse. Though, note that abstraction can still lead to this kind of application. For example, in an implementation of efficient queues by means of two lists as presented by Okasaki (1998) we have to regularly check whether the result of an application of reverse is
the empty list. In summary, to provide the best implementation we might have to evaluate which kinds of contexts are the most common ones.

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## 8 Conclusion

In the first part of this chapter we summarize the most important results and observations of the previous sections. While we already provide some ideas for future work in this first part, in the second part of this chapter we discuss possibilities for more sophisticated future work within the area of functional programming as well as an extension to the area of functional logic programming.

### 8.1 Summary

We have started by motivating why less strict functions matter. More precisely, in Section 3.2 we have considered the function intersperse from the standard Haskell library Data.List. We have observed that intersperse can cause a space leak and that this leak is due to intersperse being unnecessarily strict. By deferring a single pattern matching of intersperse, we have decreased the space usage of a program that replaces an umlaut in Shakespeare's collected works by a factor of 20,000.

Then, in Section 4.1 we have observed that the current approach to unnecessarily strict functions by Chitil (2006) proposes refactorings that have to use non-pure features like non-monadic concurrency. For example, the standard implementation of the list concatenation ( + ) is not least strict in the sense of Chitil (2006). A least strict concatenation is supposed to satisfy [1] $+\perp \equiv 1: \perp$ as well as $\perp+[1] \equiv \perp: \perp$ (Chitil 2011). The standard Haskell implementation of (\#) satisfies the former but not the latter requirement. Moreover, there is no implementation that satisfies both requirements without using non-pure features like concurrency.

Once we are convinced that we do not want to implement functions like list concatenation using non-pure features, we have to refine the notion of least strict. A function is least strict if there exists no less strict function, where the notion of function is defined in the sense of a denotational semantics, that is, we consider monotonic and continuous functions on corresponding domains. In contrast, in Section 4.2 we state that a function is minimally strict if there exists no less strict function that is also sequential. That is, we consider only functions that are monotonic, continuous, and sequential. In this way we exclude functions that have to use non-pure features.

Using this approach we define that a function is minimally strict if there exists no less strict, sequential function. In the context of sequential functions we use the term "minimally strict" and not "least strict" because, regarding sequential functions, there does not always exist a least element with respect to the less-strict ordering. For example, while the parallel and is least strict, there are two implementations of the Boolean conjunction that are minimally strict, one that performs pattern matching on its first argument, sometimes called left-strict, and one that performs pattern matching on its second argument, sometimes called right-strict.

Checking manually whether the proposed improvements with respect to strictness, like the improvements proposed for ( + ), result in a non-sequential implementation is quite challenging. Therefore, in Section 4.3 we have presented a first step towards a more detailed investigation, namely, a criterion to check whether there exists a less strict, sequential function. Moreover, we have proved that this criterion is necessary and sufficient for the existence of a less strict, sequential function. This criterion is defined for monomorphic, first-order functions, only. While we have later considered an extension to polymorphic functions we did not examine higher-order functions.

This criterion provides a basis for further investigations into checking whether functions are minimally strict. First of all, it allows us to formally argue whether a function is minimally strict by a more mechanical approach than using the definition of sequentiality. Furthermore, the criterion provides the means to develop new tools to check whether a function is minimally strict. For instance, while we have presented a tool that regards only the "semantics" of a function in form of an higher order argument, we can use the criterion to implement more sophisticated tools that incorporate the syntactical definition of a function.

As indicated, on the basis of the formal criterion we have implemented a lightweight tool, called Sloth, that checks whether a function is unnecessarily strict for arguments up to a given size. Here, size refers to the number of constructors of a term. In Chapter 5 we have presented several optimizations to a naive implementation of the tool that are used to improve the practical applicability. Nevertheless, we consider Sloth a prototype as it still has some limitations.

Most notably, testing a function that uses a more sophisticated data type can result in a quite large number of potential counter-examples that are no counter-examples if we consider arguments of larger sizes. As an extreme example consider the data type Char. Conceptually, this data type has 1114112 different constructors. The current implementation of Sloth considers only the 128 most common characters. The size, that is, the number of constructors, of any element of type Char is one. Thus, we would already generate test cases for all characters if we check a function for arguments up to size one. Therefore, we treat characters, as well as well as integers, differently. Namely, we consider these data types in the manner proposed by Runciman et al. (2008). The size of an element of type Int is the absolute value of the integer. In contrast, the size of an element of type Char is the number of bits of its integer representation. Hence, the number of test cases of type Char grows exponentially in the size while the number of test cases of type Int grows linearly in the size. Although this choice is fixed right now, it would be desirable to have control over it, for example, to be able to get an exponential growth for integers, as well.

However, this kind of enumeration of primitive data types causes other problems. Whether a given function is minimally strict or not might depend on one specific test case. Consider, for example, the function lines, which breaks up a string into a list of strings at all positions of a newline character. To observe that this function is minimally strict we have to generate a test case that contains the character ' $\backslash \mathrm{n}$ '. Depending on the order of the generation of characters, we might have to generate all other characters before we generate a newline character. For all strings that do not contain a newline character, lines yields a list with a single element. Thus, if
we only consider strings that do not contain the character ' $\backslash \mathrm{n}$ ', it looks as if lines yields a list with a single element in all cases. Therefore, we might wrongly attribute the function lines as unnecessarily strict as it could yield a list with one element without inspecting its argument. More precisely, for every possible partial value, the function lines is classified as potentially unnecessarily strict as long as we do not consider a more defined argument that contains a newline character.

For example, with the current implementation of Sloth we have to check lines for strings up to size nine to observe that the application lines $(\perp: \perp)$ is not unnecessarily strict. When we consider arguments up to size nine, we, among other test cases, generate the test case ' $\backslash \mathrm{n}$ ' $: \perp$ and, therefore, observe that the application lines $(\perp: \perp)$ is not unnecessarily strict. However, in this case, Sloth reports potential counter-examples for all applications of lines to arguments that are more defined than $\perp: \perp$, like 'a' $: \perp$, as Sloth does not generate a test case that is more defined than 'a' $: \perp$ and contains a newline character. In cases like this an interactive presentation of counter-examples, like it is used by StrictCheck (Chitil 2006), is advantageous. We have implemented a function called interactCheck that presents counter-examples one at a time and only presents the next counter-example if demanded. By employing this function we are able to only consider the most reliable counter-example. Sloth first presents the counter-examples based on the largest number of values for the corresponding infimum.

The problem of large numbers of potential counter-examples is not restricted to functions that use primitive data types like Char and Int. We can get similar behavior for functions that use ordinary algebraic data types with several constructors. To provide a more elaborated evaluation of the practical applicability of Sloth, we have to collect more data, that is, apply Sloth to more examples. Additionally, lines demonstrates another problem, we are faced with. Namely, we might have to generate a quite large number of test cases to observe that one specific application is not unnecessarily strict. Testing a function for a large number of test cases is time as well as space consuming or can even be infeasible.

To reduce the number of test cases we would like to employ background knowledge by the user. For example, to check the function lines it would be sufficient to consider strings that consist only of the undefined character $\perp$, an arbitrary defined character like 'a', and the special newline character ' $\backslash \mathrm{n}$ '. The function lines handles all other defined characters in the same way as it handles the character ' a '. With this knowledge we may define a test case generator that only generates these kinds of strings. Thus, user defined test case generators are a topic for future work. In the area of property-based testing there are even approaches to automatically reduce the number of test cases in this way, using a criterion for code coverage. In other words, a test case is not considered if there exists another test case that covers the same code with respect to some kind of criterion. For example, Gill and Runciman (2007) have presented an approach for displaying code coverage while Fischer and Kuchen (2007) use code coverage to automatically reduce the number of test cases.

In Chapter 6 we have extended the less-strict relation to polymorphic functions. Furthermore, we have presented an approach that is related to the reduction of test cases for lines, as presented above. Namely, we use additional knowledge, provided by the type of a function, to reduce the number of test cases for polymorphic func-
tions. If a function is polymorphic, it basically cannot perform pattern matching on the polymorphic component. We have proved that we can check whether a polymorphic function is minimally strict by checking its monomorphic integer instance. We don't even have to check the polymorphic function for all possible inputs of the monomorphic integer instance but only for a small number of test cases. More precisely, we have to check only a linear number of test cases in the number of polymorphic components in the data structure. For example, to test a polymorphic function that takes lists as arguments we have to check only a linear number of test cases in the length of the list. We have integrated this approach to testing polymorphic function into Sloth and we can easily check polymorphic functions efficiently using a simple type annotation.

Although, as illustrated, interpreting potential counter-examples for complex data types is problematic in some cases, even if we consider quite simple data types like lists and trees, there are lots of interesting examples of unnecessarily strict functions. As an example, the innocent looking function reverse is, somehow surprisingly, not minimally strict. Furthermore, at least we would not have guessed that there are any unnecessarily strict functions in a standard Haskell library like Data.List or that a well-known function like Peano multiplication is unnecessarily strict. Furthermore, in Section 7.4 we have illustrated that we can use Sloth to improve the memory behavior of a real world Haskell package, namely, the package split. Finally, in Section 7.5, we have illustrated that we can always derive a context in which an unnecessarily strict implementation performs worse than the corresponding less strict implementation. This result implies that a less strict implementation is not necessarily better than the original implementation but it is also not worse. All in all, a prototype implementation like Sloth is quite useful for collecting examples of unnecessarily strict functions as we have presented throughout this work.

### 8.2 Future Work

In this section we present future work, first, in the area of functional programming and, then, in the area of functional logic programming. Note that we do not repeat all statements about future work that have been mentioned in the previous chapters.

### 8.2.1 Functional Programming

To motivate one direction for future work let us consider a case study by Runciman and Wakeling (1993b) that is "famous" for showing that some space leaks can be fixed by using an implementation that is "more lazy" (Hudak et al. 2007). We have only touched this particular example in Chapter 1 as the notion of "more lazy" with respect to this example is more sophisticated than our notion of less strict. Runciman and Wakeling (1993b) have presented a tool for profiling the memory usage of programs in a lazy functional programming language. Using this tool, they have investigated the memory usage of a program that takes a propositional formula and yields a clausal normal form of this formula. They have observed that the memory usage can be improved by using a "more lazy" implementation of one particular
function. However, in the sense of the less-strict relation, presented here, the proposed improved implementation is incomparable to the original implementation. The function in question yields a list and, with respect to some arguments, the improved implementation changes the order of the elements of the list in comparison to the original implementation. Therefore, these functions are incomparable with respect to the less-strict relation. As the algorithm employs lists as a simple implementation of sets, the order of the elements does not matter. This example demonstrates that we sometimes do not look for a less strict implementation with respect to the canonical semantic ordering that is based on the structure of the result data type but for a less strict implementation with respect to a more abstract ordering. For example, in the case at hand we would like to use an ordering that ignores the order of the elements in the list. This example can be considered as one instance of the more general case of testing whether a function that uses an abstract data type is minimally strict.

To illustrate another related limitation of the less-strict relation we consider the purely functional implementation of a certain on-line algorithm by Bird et al. (1997). By employing a less strict matrix transposition they were able to provide an implementation of the considered on-line algorithm that has linear complexity. In this case a matrix is represented by a list of lists. If we check the original, unnecessarily strict implementation of the transposition, Sloth does not report any counter-examples. In fact, the standard implementation as well as the implementation presented by Bird et al. (1997) are minimally strict. But, again, these functions are incomparable as they yield incomparable results for total arguments. More precisely, these function do not agree for arguments that are outside of the intended domain of the functions. A transposition that takes a list of lists as argument is only well defined if we apply it to a list that contains lists of equal length, as other kinds of lists of lists do not represent valid matrices. We conjecture that these two implementations of transposition become comparable with respect to the less-strict relation, if we consider only valid inputs. Hence, for future work we would like to incorporate a means into Sloth to constrain the considered arguments of a function. A naive way to constrain the arguments is to apply a partial identity to the arguments of a function and map all undesired arguments to $\perp$. However, it is very difficult to implement the corresponding partial identity as we have to regard partial values.
In our opinion, the most important and probably most difficult future work is a connection between the less-strict relation and space usage. As we have illustrated in Section 7.5, we can always construct a context in which a less strict implementation is an improvement over an unnecessarily strict implementation. As a first step, we have to formally verify this informally stated conjecture. Furthermore, we have to verify a second conjecture that is concerned with reverse. We assume that there is no minimally strict implementation of reverse that is an improvement over the unnecessarily strict implementation with respect to all possible contexts. If we can show that there are functions for which no minimally strict function exists that is an improvement with respect to all contexts, we provide a strong argument for considering only functions that are minimally strict with respect to a specific context.

A syntax-driven approach by means of transformations like case deferment might provide an easier access to a connection between memory consumption and strict-
ness. It is much easier to reason about memory behavior of two functions that are related by a transformation like case deferment than reasoning about memory behavior of two arbitrary functions that are related by the less-strict relation. If we consider two arbitrary functions that are related by the less-strict relation the functions can behave quite differently from an operational point of view while two functions that are related by the case deferment behave operationally quite similar.

Furthermore, we would like to take a closer look at the trade-off between abstraction and minimally strict implementations. In Section 7.3 we have already observed, using the example of the function $(<=)$, that abstraction by means of a default implementation of a type class function can result in an unnecessarily strict implementation. As another example for the trade-off between modularity and a minimally strict implementation, consider the function sequence :: Monad $m \Rightarrow[m \alpha] \rightarrow m[\alpha]$. If we consider the list monad instance of this function, it calculates the cross product of an arbitrary number of lists. For example, we have

$$
\text { sequence }[[1],[3,4],[5,6]] \equiv[[1,3,5],[1,3,6],[1,4,5],[1,4,6]] \text {. }
$$

Because sequence is implemented generically for all monads, its implementation for the list mond is unnecessarily strict. For example, the evaluation of the expression sequence [[1..],[]] does not terminate. In contrast, a minimally strict implementation would yield the empty list instead. Simplified, the reason is that the list instance of sequence can check whether any for the elements in the argument list is the empty list while an implementation for an arbitrary monad cannot.

Finally, we could take a quite contrary approach in comparison to the basic idea of minimally strict functions. Instead of considering sequential functions only, we can stick to the concept of least strict functions and consider a more powerful language that makes it possible to define least strict functions. For example, we can employ the Haskell operator unamb by Elliott (2009), which is an implementation of the quite intensively studied operator amb by McCarthy (1961). However, we assume that this operator causes a significant run-time overhead that cannot be outweighed by the benefits of a least strict definition.

Even more revolutionary, we would like to investigate a more declarative approach to pattern matching in lazy functional programming. As we have touched on in the introduction the current approach to pattern matching in Haskell is rather non-declarative as a programmer specifies the order of evaluation by means of case expressions ${ }^{1}$. More precisely, we want to regard the definition of pattern matching only with respect to total values and provide a transformation that looks for the optimal implementation with respect to partial values. In this context we have to mention that there are even more declarative approaches to the transformation of rules into case expressions. More precisely, the programming language Curry uses a transformation (Antoy 1992) that can be regarded as more declarative as it abstract over the order of the rules and results in less strict functions.

[^17]
### 8.2.2 Functional-Logic Programming

We expect that non-strict functional logic programming languages like Curry are a very interesting field of future work. As we have illustrated in Section 7.3, deterministic functions that are applied to free variables can benefit from less strict implementations. But, we can also consider non-deterministic functions. To be applicable to full Curry we would have generalize the notation of minimally strict functions to non-deterministic functions. We have already presented a denotational semantics for Curry (Christiansen et al. 2011a) ${ }^{2}$ that seems adequate for further investigations as it provides a kind of functional view on functional logic programming and, therefore, should allow for a quite natural extension of the less strict relation to non-deterministic functions.

To illustrate that the extension to non-deterministic functions is valuable let us consider the following function, called insertND, that non-deterministically inserts an element at any position of a list. The operator (?) denotes a non-deterministic choice of its two arguments. That is, if the argument of insertND is a non-empty list $y: y s$, we either insert the element $x$ at the front of $y: y s$ or we insert it non-deterministically in the list $y s$ and put $y$ to the front of the result.

```
insertND :: \(\alpha \rightarrow[\alpha] \rightarrow[\alpha]\)
insertND \(x[] \quad=[x]\)
insertND \(x(y: y s)=(x: y: y s) ?(y: \operatorname{insertND} x y s)\)
```

This is a standard implementation of the non-deterministic insertion function that, for example, is proposed on a webpage with Curry example programs (Hanus).

However, there is another implementation of this function that behaves equally for all total arguments. As overlapping rules induce a non-deterministic choice, we can define one rule that matches in the case that the list is empty as well as in the case that the list is non-empty as follows.

$$
\begin{aligned}
& \operatorname{insertND^{\prime }::\alpha \rightarrow [\alpha ]} \rightarrow[\alpha] \\
& \text { insertND } x \text { ys }=x: y s \\
& \text { insertND' } x(y: y s)=y: \text { insertND } x \text { ys }
\end{aligned}
$$

The function insert $N D^{\prime}$ is less strict than insertND if we consider a less-strict relation that is based on the semantic ordering used to model a denotational semantics for a non-deterministic language (Christiansen et al. 2011a). For example, we have insertND $x \perp \equiv \perp$ but insertND $x \perp \equiv x: \perp$. While in this case the result of both functions is actually kind of deterministic, there are also non-deterministic cases. For example, we have insertND $x(y: \perp) \equiv(x: y: \perp) ?(y: \perp)$ and insertND $D^{\prime} x(y: \perp) \equiv(x: y: \perp) ?(y: x: \perp)$. Therefore, to state that insertND' is less strict than insertND we have to regard non-determinism.

Let us consider an implementation of permutation sort, which is a standard example for functional logic programming. If we implement permutation sort by means of these functions we get a much more efficient implementation using the "less strict" implementation of insertND. For example, it takes around 20 times as

[^18]
## 8 Conclusion

long to sort a list with 15 elements using insertND than it takes to perform the same task with insertND'.

Furthermore, we can even observe another close relation between this example and a result presented in this thesis. We can derive the implementation insertND ${ }^{\prime}$ from the implementation insertND by means of a generalization of case deferment to functional logic programming. More precisely, if we consider an equivalent implementation of insertND that is defined by means of (?), we can derive insert $N D^{\prime}$ from insertND by moving the context $(x: x s)$ ? [•] over a case expression.

Finally, we have also started to investigate free theorems in a functional logic programming language (Christiansen et al. 2010). More precisely, we have provided an example-driven investigation that shows how we can adapt free theorems to the context of functional logic programming. A formalization by means of the semantics presented in (Christiansen et al. 2011a) will allow us to generalize the observations from Chapter 6 to functional logic languages.

## A Proofs from Chapter 4

## A. 1 Proofs from Section 4.2

Proof (of Lemma 4.2.4): We prove this statement by structural induction over the typing rules.

Base Cases: Obviously we have $\left(\llbracket e \rrbracket_{a_{1}}^{E r r}, \llbracket e \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau}^{m}$ for the rules $\Gamma, x:: \tau \vdash x:: \tau$, $\Gamma \vdash$ True $::$ Bool, $\Gamma \vdash$ False $::$ Bool, $\Gamma \vdash$ Nil $_{\tau}::$ List $^{\tau}$, and $\Gamma \vdash$ undefined $_{\tau}:: \tau$.

Inductive Cases: We consider the rule for list construction.

$$
\frac{\Gamma \vdash e_{1}:: \tau \quad \Gamma \vdash e_{2}:: \text { List } \tau}{\Gamma \vdash \operatorname{Cons}\left\langle e_{1}, e_{2}\right\rangle:: \text { List } \tau}
$$

By induction hypothesis we have $\left(\llbracket e_{1} \rrbracket_{a_{1}}^{E r r}, \llbracket e_{1} \rrbracket_{a_{2}}\right) \in \operatorname{err} \tau_{\tau}^{m}$ and $\left(\llbracket e_{2} \rrbracket_{a_{1}}^{E r r}, \llbracket e_{2} \rrbracket_{a_{2}}\right) \in$ err ${ }_{\text {List } \tau}^{m}$. We reason as follows.

$$
\begin{aligned}
&\left(\llbracket e_{1} \rrbracket_{a_{1}}^{E r r}, \llbracket e_{1} \rrbracket \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau}^{m} \wedge\left(\llbracket e_{2} \rrbracket_{a_{1}}^{E r r}, \llbracket e_{2} \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\text {List }}^{m} \\
& \Longleftrightarrow\left(\operatorname{Cons}\left\langle\llbracket e_{1} \rrbracket_{a_{1}}^{E r r}, \llbracket e_{2} \rrbracket_{a_{1}}^{E r r}\right\rangle, \operatorname{Cons}\left\langle\llbracket e_{1} \rrbracket_{a_{1}}, \llbracket e_{2} \rrbracket_{a_{1}}\right\rangle\right) \in \operatorname{crr}_{\text {List }}^{m} \\
& \Longleftrightarrow\left(\operatorname{Cons} \llbracket\left\langle e_{1}, e_{2}\right\rangle \rrbracket_{a_{1}}^{E r}, \operatorname{Cons} \llbracket\left\langle e_{1}, e_{2}\right\rangle \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\text {List } \tau}^{m} \\
& \Longleftrightarrow\left(\llbracket \operatorname{Cons}\left\langle e_{1}, e_{2}\right\rangle \rrbracket_{a_{1}}^{E r r}, \llbracket \operatorname{Cons}\left\langle e_{1}, e_{2}\right\rangle \rrbracket_{a_{2}}\right) \in \text { err }_{\text {List } \tau}^{m}
\end{aligned}
$$

Next we consider the rule for tuple construction.

$$
\frac{\Gamma \vdash e_{1}:: \tau_{1} \quad \ldots \quad \Gamma \vdash e_{n}:: \tau_{n}}{\Gamma \vdash\left\langle e_{1}, \ldots, e_{n}\right\rangle:: \tau_{1} \times \cdots \times \tau_{n}}
$$

We have $\left(\llbracket e_{i} \rrbracket_{a_{1}}^{E r r}, \llbracket e_{i} \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau_{i}}^{m}$ for all $i \in\{1, \ldots, n\}$ by induction hypothesis and reason as follows.

$$
\begin{aligned}
& \forall i \in\{1, \ldots, n\} .\left(\llbracket e_{i} \|_{a_{1}}^{E r r}, \llbracket e_{i} \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau_{i}}^{m} \\
& \Longleftrightarrow\left(\llbracket\left\langle e_{1}, \ldots, e_{n}\right\rangle \rrbracket_{a_{1}}^{E r r}, \llbracket\left\langle e_{1}, \ldots, e_{n}\right\rangle \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau_{1} \times \cdots \times \tau_{n}}^{m}
\end{aligned}
$$

In the case of a function application we have to consider the following rule.

$$
\frac{\Gamma \vdash f:: \sigma \rightarrow \tau \quad \Gamma \vdash e:: \sigma}{\Gamma \vdash f e:: \sigma}
$$

## A Proofs from Chapter 4

By induction hypothesis we have $\left(\llbracket f \rrbracket_{a_{1}}^{E r r}, \llbracket f \rrbracket_{a_{2}}\right) \in \operatorname{err} r_{\sigma \rightarrow \tau}^{m}$ and $\left(\llbracket e \rrbracket_{a_{1}}^{E r r}, \llbracket e \rrbracket_{a_{2}}\right) \in$ $e r r_{\sigma}^{m}$ and reason as follows.

$$
\begin{aligned}
& \left(\llbracket f \rrbracket_{a_{1}}^{E r r}, \llbracket f \rrbracket_{a_{2}}\right) \in \operatorname{err}{ }_{\sigma \rightarrow \tau}^{m} \wedge\left(\llbracket \ell \rrbracket_{a_{1}}^{E r r}, \llbracket e \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\sigma}^{m} \\
& \Longleftrightarrow \forall(a, b) \in \operatorname{err}_{a}^{m} .\left(\llbracket f \rrbracket_{a_{1}}^{E r r} a, \llbracket f \rrbracket_{a_{2}} b\right) \in \operatorname{err}_{\tau}^{m} \wedge\left(\llbracket \ell \rrbracket_{a_{1}}^{E r r}, \llbracket \ell \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\sigma}^{m} \\
& \Longrightarrow\left(\llbracket f \rrbracket_{a_{1}}^{E r r} \llbracket \rrbracket \rrbracket_{a_{1}}^{E r r}, \llbracket f \rrbracket_{a_{1}} \llbracket e \rrbracket_{a_{2}}\right) \in e r r_{\tau}^{m} \\
& \Longrightarrow\left(\llbracket f e \rrbracket_{a_{1}}^{E r r}, \llbracket f e \rrbracket_{a_{2}}\right) \in e r r_{\tau}^{m}
\end{aligned}
$$

We consider the rule for function types.

$$
\frac{f:: \sigma \rightarrow \tau \in P}{\Gamma \vdash f:: \sigma \rightarrow \tau}
$$

As the program $P$ is well-typed we have $\Gamma \vdash f:: \sigma \rightarrow \tau$. Therefore, by induction hypothesis we have $\left(\llbracket f \rrbracket_{a_{1}}^{E r r}, \llbracket f \rrbracket_{a_{2}}\right) \in \operatorname{err} r_{\sigma \rightarrow \tau}^{m}$.
We consider the rule for list case expressions.

$$
\frac{\Gamma \vdash e:: \text { List } \tau \quad \Gamma \vdash e_{1}:: \tau^{\prime} \quad \Gamma, x:: \tau, x \mathrm{~s}:: \text { List } \tau \vdash e_{2}:: \tau^{\prime}}{\Gamma \vdash \text { case } e \text { of }\left\{\mathrm{Nil}_{\tau} \rightarrow e_{1} ; \operatorname{Cons}\langle x, x s\rangle \rightarrow e_{2}\right\}:: \tau^{\prime}}
$$

We have $\left(\llbracket e \rrbracket_{a_{1}}^{E r r}, \llbracket e \rrbracket_{a_{2}}\right) \in \operatorname{err} r_{\text {List } \tau}^{m}$ by induction hypothesis and distinguish five cases.

- If $\llbracket e \rrbracket_{a_{1}}^{E r r}=\perp_{\text {List } \tau}$, by definition of case we have to show that

$$
\left(\perp_{\tau^{\prime}}, \llbracket \text { case } e \text { of } \ldots \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau^{\prime}}^{m}
$$

This statement is true because for all types $\tau$ we have $\left(\perp_{\tau}, v\right) \in \operatorname{err}_{\tau}^{m}$ for all $v \in \llbracket \tau \rrbracket$.

- If $\llbracket e \rrbracket_{a_{1}}^{E r r}=E r r o r_{\text {List } \tau} n$ and $m \neq n$ by definition of case we have to show that

$$
\left(\text { Error }_{\tau^{\prime}} n, \llbracket \text { case } e \text { of } \ldots \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau^{\prime}}^{m}
$$

If $m \neq n$ for all types $\tau$ we have $\left(\right.$ Error $\left._{\tau} n, v\right) \in \operatorname{err}_{\tau}^{m}$ for all $v \in \llbracket \tau \rrbracket$.

- If $\llbracket e \rrbracket_{a_{1}}^{E r r}=E r r o r_{\text {List } \tau} m$ and $\llbracket e \rrbracket_{a_{2}}=\perp_{\text {List } \tau}$ by definition of case we have to show that $\left(E r r o r_{\tau^{\prime}} m, \perp_{\tau^{\prime}}\right) \in$ err $_{\tau^{\prime}}^{m}$ which is true for all types $\tau^{\prime}$.
- If $\llbracket e \rrbracket \rrbracket_{a_{1}}^{E r r}=N i l_{\tau}$ and $\llbracket e \rrbracket_{a_{2}}=N i l_{\tau}$ by definition of case we have to show that $\left(\llbracket e_{1} \rrbracket_{a_{1}}^{E r r}, \llbracket e_{1} \rrbracket \rrbracket_{a_{2}}\right) \in \operatorname{err}_{\tau^{\prime}}^{m}$ which is given by induction hypothesis.
- If $\llbracket e \rrbracket_{a_{1}}^{E r r}=\mathrm{Cons}\left\langle v_{1}, v_{2}\right\rangle$ and $\llbracket e \rrbracket_{a_{2}}=$ Cons $\left\langle v_{3}, v_{4}\right\rangle$ with $\left(v_{1}, v_{3}\right) \in \operatorname{err}{ }_{\tau}^{m}$ and $\left(v_{2}, v_{4}\right) \in \operatorname{err}$ List $\tau$ by definition of case we have to show that

$$
\left(\llbracket e_{2} \rrbracket_{a_{1}\left[x \mapsto v_{1}, x s \mapsto v_{2}\right]}^{E r r}\left[\llbracket e_{2} \rrbracket_{a_{2}\left[x \mapsto v_{3}, x s \mapsto v_{4}\right]}\right) \in \operatorname{err}_{\tau^{\prime}}^{m}\right.
$$

which is given by induction hypothesis.

The proof for Boolean case expressions is analogous to the proof for list case expressions.

Proof (of Lemma 4.2.5): Because $p$ is a demanded position in $p v$ at position $r p$, there exists $e v \in \llbracket \sigma \rrbracket^{E r r}$ such that $(e v, p v) \in e r r_{\sigma}^{n}$. Furthermore, position $p$ is the only position such that

$$
\left.e v\right|_{p}=\text { Error } n
$$

and

$$
\left.\left(\llbracket f \rrbracket^{E r r} e v\right)\right|_{r p}=\text { Error } n .
$$

To show that $p$ is a sequential position in $p v$ at position $r p$ with respect to $\llbracket f \rrbracket$ we first have to show that $\left.(\llbracket f \rrbracket p v)\right|_{r p}=\perp$. We have $(e v, p v) \in e r r_{\sigma}^{n}$, and by Lemma 4.2.4 we get $\left(\llbracket f \rrbracket^{E r r}, \llbracket f \rrbracket\right) \in e r r_{\tau}^{n}$ and, thus, $\left(\llbracket f \rrbracket^{E r r} e v, \llbracket f \rrbracket p v\right) \in \operatorname{err}_{\tau}^{n}$. By definition of demanded positions we get

$$
\left.\left(\llbracket f \rrbracket^{E r r} e v\right)\right|_{r p}=\text { Error } n
$$

and get

$$
\left.(\llbracket f \rrbracket p v)\right|_{r p}=\perp
$$

because $\left(\llbracket f \rrbracket^{E r r} e v, \llbracket f \rrbracket p v\right) \in e r r_{\tau}^{n}$.
Finally, we have to show that for all $p v^{\prime} \in \llbracket \sigma \rrbracket$ the following holds.

$$
\left.p v^{\prime} \sqsupseteq p v \wedge p v^{\prime}\right|_{p}=\left.\perp \Longrightarrow(\llbracket f \rrbracket p v)\right|_{r p}=\perp
$$

Let $p v^{\prime} \in \llbracket \sigma \rrbracket$ such that $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p}=\perp$. As position $p$ of $e v$ contains an exception, that is, $\left.e v\right|_{p}=$ Error $n$, by definition of demanded positions, $p$ is the only position that contains the error labeled with $n$. This implies $\left(e v, p v^{\prime}\right) \in e r r_{\sigma}^{n}$. By Lemma 4.2.4 we have $\left(\llbracket f \rrbracket^{E r r}, \llbracket f \rrbracket\right) \in e r r_{\tau}^{n}$ and, thus, $\left(\llbracket f \rrbracket^{E r r} e v, \llbracket f \rrbracket p v^{\prime}\right) \in e r r^{n}$. Furthermore, as

$$
\left.\left(\llbracket f \rrbracket^{E r r} e v\right)\right|_{r p}=\text { Error } n
$$

we get

$$
\left.\left(\llbracket f \rrbracket p v^{\prime}\right)\right|_{r p}=\perp
$$

because $\left(\llbracket f \rrbracket^{E r r} e v, \llbracket f \rrbracket p v^{\prime}\right) \in e r r_{\tau}^{n}$. This finally completes the proof that $p$ is a sequential position in $p v$ at position $r p$.

## A. 2 Proofs from Section 4.3.1

Proof (of Lemma 4.3.1): As $f$ is sequential, there exists a sequential position $p^{\prime}$ in $v$ at result position $r p$. We have $p^{\prime} \not \leq p$ and $p \not \leq p^{\prime}$. By definition of sequential position we have $\left.v\right|_{p^{\prime}}=\perp$ and, therefore, $\left.\left(v\left[v^{\prime}\right]_{p}\right)\right|_{p^{\prime}}=\left.v\right|_{p^{\prime}}=\perp$ by Lemma 4.2.2. Finally, because $v\left[v^{\prime}\right]_{p} \sqsupseteq v[\perp]_{p}=v$ and $p^{\prime}$ is a sequential position in $v$ we get $\left.\left(f v\left[v^{\prime}\right]_{p}\right)\right|_{r p}=$ $\perp$.

## A Proofs from Chapter 4

Proof (of Lemma 4.3.2): Because $v \neq \perp$, there exists $v^{\prime} \in \llbracket \tau \rrbracket$ and $p \in \operatorname{Pos} v^{\prime}$ such that $v^{\prime} \lessdot_{p} v$. By monotonicity there exists $r p^{\prime} \in \operatorname{Pos}\left(f v^{\prime}\right)$ such that $\left.\left(f v^{\prime}\right)\right|_{r p^{\prime}}=\perp$ and $r p^{\prime} \leq r p$. Because we have $\left.\left(f v^{\prime}\right)\right|_{r p^{\prime}}=\perp$, we have either $v^{\prime} \rightsquigarrow_{r p^{\prime}} v$ or $v^{\prime} \mapsto_{r p^{\prime}} v$. If we have a non-sequential step $v^{\prime} \rightsquigarrow_{r p^{\prime}} v$, by Lemma 4.3.1 we have $\left.(f v)\right|_{r p^{\prime}}=\perp$, and by definition of $\sqsubseteq$ we get $r p=r p^{\prime}$.

Lemma A.2.1: If we have $v_{1} \rightsquigarrow r p_{1} v_{2}$ and $\left.\left(f v_{2}\right)\right|_{r p_{2}}=\perp$ with $r p_{1} \leq r p_{2}$, then $r p_{1}=r p_{2}$.

Proof: We have $v_{2}=v_{1}[C \perp]_{p}$ for some constructor $C$ and a non-sequential position $p$. By Lemma 4.3.1 we have $\left.\left(f v_{2}\right)\right|_{r p_{1}}=\left.\left(f v_{1}[C \perp]_{p}\right)\right|_{r p_{1}}=\perp$. By definition of $\sqsubseteq$ we get $r p_{1}=r p_{2}$.

Lemma A.2.2: If $v_{1} \rightsquigarrow_{r p} v_{2} \hookrightarrow_{r p} v_{3}$ and $v_{1} \lessdot_{p} v_{2} \lessdot_{p^{\prime}} v_{3}$ then $p \not \leq p^{\prime}$.
Proof: As $v_{1} \rightsquigarrow_{r p} v_{2}$ there exists $v \in \llbracket \tau \rrbracket$ with $v \sqsupseteq v_{1},\left.v\right|_{p}=\perp$ and $\left.(f v)\right|_{r p} \neq \perp$. We have $v=v[\perp]_{p} \sqsubseteq v\left[\left.v_{2}\right|_{p}\right]_{p}$ by monotonicity of $v[\cdot]_{p}$ and by monotonicity of $f$ we have

$$
\perp \neq\left.\left.(f v)\right|_{r p} \sqsubseteq\left(f v\left[\left.v_{2}\right|_{p}\right]_{p}\right)\right|_{r p}
$$

that is, $\left.\left(f v\left[\left.v_{2}\right|_{p}\right]_{p}\right)\right|_{r p} \neq \perp$. By monotonicity of $\cdot\left[\left.v_{2}\right|_{p}\right]_{p}$ we have $v_{2}=v_{1}\left[\left.v_{2}\right|_{p}\right]_{p} \sqsubseteq$ $v\left[\left.v_{2}\right|_{p}\right]_{p}$. If $p \leq p^{\prime}$ we have $\left.v\left[\left.v_{2}\right|_{p}\right]_{p}\right|_{p^{\prime}}=\left.v_{2}\right|_{p^{\prime}}$ by Lemma 4.2.2 and, therefore, $p^{\prime}$ is a non-sequential position in $v_{2}$ at position $r p$. This is a contradiction as $p^{\prime}$ is a sequential position in $v_{2}$ at position $r p$. That is, we have $p \not \leq p^{\prime}$.

Proof (of Lemma 4.3.3): There exist positions $p$ and $p^{\prime}$ such that $v_{1} \lessdot p v_{2}$ and $v_{2} \lessdot p^{\prime}$ $v_{3}$. Because we have $v_{1} \lessdot_{p} v_{2}$ there exists a constructor $C$ such that $\left.v_{2}\right|_{p}=C \perp$ and $\left.v_{1}\right|_{p}=\perp$. Because we have $v_{2} \lessdot_{p^{\prime}} v_{3}$ there exists a constructor $D$ such that $\left.v_{3}\right|_{p^{\prime}}=D \perp$ and $\left.v_{2}\right|_{p^{\prime}}=\perp$. By Lemma A.2.2 we have $p \not \leq p^{\prime}$ and $p^{\prime} \not \leq p$ and set $v:=v_{1}[D \perp]_{p^{\prime}}$. We get $v[C \perp]_{p}=v_{3}$ and, thus, $v_{1} \lessdot_{p^{\prime}} v$ and $v \lessdot_{p} v_{3}$.

We have to show that $p^{\prime}$ is a sequential position in $v_{1}$ at position $r p_{1}$. That is, we have to show that for all $v^{\prime} \in \llbracket \tau \rrbracket$ the following holds.

$$
\left.v^{\prime} \sqsupseteq v_{1} \wedge v^{\prime}\right|_{p^{\prime}}=\left.\perp \Longrightarrow\left(f v^{\prime}\right)\right|_{r p_{1}}=\perp
$$

Let $v^{\prime} \in \llbracket \tau \rrbracket$ with $v^{\prime} \sqsupseteq v_{1}$ and $\left.v^{\prime}\right|_{p^{\prime}}=\perp$. We distinguish two cases.
Case $1\left(v^{\prime} \sqsupseteq v_{2}\right)$ : Because $v^{\prime} \sqsupseteq v_{2},\left.v^{\prime}\right|_{p^{\prime}}=\perp$, and $p^{\prime}$ is a sequential position in $v_{2}$ at position $r p_{1}$, we have $\left.\left(f v^{\prime}\right)\right|_{r p_{1}}=\perp$.
Case $2\left(v^{\prime} \nsupseteq v_{2}\right)$ : We consider the values $v_{1}\left[\left.v^{\prime}\right|_{p}\right]_{p}$ and $v^{\prime}[C \perp]_{p}$. We have

$$
v_{1}=v_{1}[\perp]_{p} \sqsubseteq v_{1}\left[\left.v^{\prime}\right|_{p}\right]_{p}
$$

by monotonicity of $v_{1}[\cdot]_{p}$,

$$
v_{1}\left[\left.v^{\prime}\right|_{p}\right]_{p} \sqsubseteq v^{\prime}\left[\left.v^{\prime}\right|_{p}\right]_{p}=v^{\prime}
$$

by monotonicity of $\cdot\left[\left.v^{\prime}\right|_{p}\right]_{p}$, and

$$
v_{2}=v_{1}[C \perp]_{p} \sqsubseteq v^{\prime}[C \perp]_{p}
$$

by monotonicity of $\cdot[C \perp]_{p}$. The following Hasse diagram depicts the situation.


In the following we consider the value $v^{\prime}[\perp]_{p}$, which is both less equal $v^{\prime}[C \perp]_{p}$ and less equal $v^{\prime}$. As $p \not \leq p^{\prime}$ and $p^{\prime} \not \leq p$ we have $\left.\left(v^{\prime}[C \perp]_{p}\right)\right|_{p^{\prime}}=\left.v^{\prime}\right|_{p^{\prime}}$ by Lemma 4.2.2. Because we have $v^{\prime}[C \perp]_{p} \sqsupseteq v_{2}$ and $p^{\prime}$ is a sequential position in $v_{2}$ we get $\left.\left(f v^{\prime}[C \perp]_{p}\right)\right|_{r p_{1}}=\perp$. Furthermore, by monotonicity we get $\left.\left(f v^{\prime}[\perp]_{p}\right)\right|_{r p_{1}}=\perp$ as well.
Next we show that we can decompose $v^{\prime}$ into $v^{\prime}[\perp]_{p}$ and $v_{1}\left[\left.v^{\prime}\right|_{p}\right]_{p}$.

$$
\begin{aligned}
v^{\prime}[\perp]_{p} \sqcup v_{1}\left[\left.v^{\prime}\right|_{p}\right]_{p} & =\left(v^{\prime} \sqcup v_{1}\right)\left[\left.\perp \sqcup v^{\prime}\right|_{p}\right]_{p} & & \text { continuity of } \cdot[\cdot]_{p} \\
& =v^{\prime}\left[\left.v^{\prime}\right|_{p}\right]_{p} & & v_{1} \sqsubseteq v^{\prime} \text { and } \perp \sqsubseteq v^{\prime} \\
& =v^{\prime} & & \text { property of }\left.\cdot\right|_{p} \text { and } \cdot[\cdot]_{p}
\end{aligned}
$$

We use this decomposition and continuity to show that we have $\left.\left(f v^{\prime}\right)\right|_{r p_{1}}=\perp$.

$$
\begin{aligned}
\left.\left(f v^{\prime}\right)\right|_{r p_{1}} & =\left.\left(f\left(v^{\prime}[\perp]_{p} \sqcup v_{1}\left[\left.v^{\prime}\right|_{p}\right]_{p}\right)\right)\right|_{r p_{1}} & & \text { decomposition } \\
& =\left.\left(f v^{\prime}[\perp]_{p} \sqcup f v_{1}\left[\left.v^{\prime}\right|_{p}\right]_{p}\right)\right|_{r p_{1}} & & \text { continuity of } f \\
& =\left.\left(f v^{\prime}[\perp]_{p}\right)\right|_{r p_{1}} \sqcup\left(\left.f\left(v_{1}\left[\left.v^{\prime}\right|_{p}\right]_{p}\right)\right|_{r p_{1}}\right. & & \text { continuity of }\left.\cdot\right|_{r p_{1}} \\
& =\left.\left(f v^{\prime}[\perp]_{p}\right)\right|_{r p_{1}} \sqcup \perp & & \text { Lemma 4.3.1 } \\
& =\perp \sqcup \perp & & \text { argumentation above } \\
& =\perp & &
\end{aligned}
$$

This shows that $p^{\prime}$ is a sequential position in $v_{1}$, that is, $v_{1} \hookrightarrow_{r p_{1}} v$.
We finally have to show that either $v \rightsquigarrow_{r p_{2}} v_{3}$ or there exists $r p_{2}$ with $r p_{1} \leq$ $r p_{2} \leq r p_{3}$ and $v \not \mapsto_{r p_{2}} v_{3}$. Because we have $\left.\left(f v_{1}\right)\right|_{r p_{1}}=\perp$ and $\left.\left(f v_{3}\right)\right|_{r p_{3}}=\perp$ by monotonicity there exists $r p_{2}$ with $r p_{1} \leq r p_{2} \leq r p_{3}$ and $\left.(f v)\right|_{r p_{2}}=\perp$. This implies either $v \rightsquigarrow r p_{2} v_{3}$ or $v \nrightarrow r p_{2} v_{3}$. By Lemma A.2.1 we have $r p_{2}=r p_{3}$ if $v \rightsquigarrow_{r p_{2}} v_{3}$.

## A Proofs from Chapter 4

Proof (of Lemma 4.3.4): By induction over $n$ we show that for all $n \in \mathbb{N}_{0}$ if

$$
v_{1}\left(\rightsquigarrow r p_{1}\right)^{n} v_{2} \mapsto_{r p_{1}} v_{3}
$$

then

$$
v_{1} \mapsto_{r p} v \mapsto^{*} v^{\prime}\left(\rightsquigarrow_{r p^{\prime}}\right)^{*} v_{3} .
$$

In the induction hypothesis we assume that the statement holds for all $m \in \mathbb{N}_{0}$ with $m \leq n$.

Base Case: We have $v_{2} \hookrightarrow_{r p_{1}} v_{3}$ and set $v:=v_{3}$ and $v^{\prime}:=v_{3}$.
Inductive Case: We have $v_{1}\left(\rightsquigarrow{ }_{r p_{1}}\right)^{n+1} v_{2} \mapsto{ }_{r p_{1}} v_{3}$. There exists $v \in \llbracket \tau \rrbracket$ such that $v_{1}\left(\rightsquigarrow r p_{1}\right)^{n} v \rightsquigarrow_{r p_{1}} v_{2} \mapsto_{r p_{1}} v_{3}$. We consider the sub-sequence $v \rightsquigarrow r p_{1} v_{2} \mapsto_{r p_{1}}$ $v_{3}$. By Lemma 4.3.3 there exists $v_{2}^{\prime} \in \llbracket \tau \rrbracket$ such that $v \mapsto_{r p_{1}} v_{2}^{\prime} \rightsquigarrow r p_{3} v_{3}$ or there exists $r p_{2}$ with $r p_{1} \leq r p_{2} \leq r p_{3}$ such that $v \mapsto r p_{1} v_{2}^{\prime} \mapsto r p_{2} v_{3}$. We distinguish these two cases.

Case $1\left(v \mapsto_{r p_{1}} v_{2}^{\prime} \rightsquigarrow r p_{3} v_{3}\right):$ We have a sequence of the form

$$
v_{1}\left(\rightsquigarrow r p_{1}\right)^{n} v \mapsto_{r p_{1}} v_{2}^{\prime} .
$$

As we have $\left.\left(f v_{2}^{\prime}\right)\right|_{r p_{3}}=\perp$ by induction hypothesis there exist $v^{\prime}, v^{\prime \prime} \in \llbracket \tau \rrbracket$ such that

$$
v_{1} \mapsto_{r p_{1}} v^{\prime} \mapsto^{*} v^{\prime \prime}\left(\rightsquigarrow r p_{3}\right)^{*} v_{2}^{\prime},
$$

which is montone with respect to result positions. Finally we get the monotonic sequence

$$
v_{1} \mapsto_{r p_{1}} v^{\prime} \mapsto^{*} v^{\prime \prime}\left(\rightsquigarrow_{r p_{3}}\right)^{*} v_{3} .
$$

Case $2\left(v \hookrightarrow r p_{1} v_{2}^{\prime} \mapsto_{r p_{2}} v_{3}\right):$ We have a sequence of the form

$$
v_{1}\left(\rightsquigarrow r p_{1}\right)^{n} v \mapsto_{r p_{1}} v_{2}^{\prime} .
$$

As we have $\left.\left(f v_{2}^{\prime}\right)\right|_{r_{p_{2}}}=\perp$ by induction hypothesis there exist $v^{\prime}, v^{\prime \prime} \in \llbracket \tau \rrbracket$ such that

$$
v_{1} \mapsto_{r p_{1}} v^{\prime} \hookrightarrow^{*} v^{\prime \prime}\left(\rightsquigarrow_{r p_{2}}\right)^{*} v_{2}^{\prime}
$$

which is montone with respect to result positions. Furthermore, there exists $m \in \mathbb{N}_{0}$ such that

$$
v^{\prime \prime}\left(\rightsquigarrow r p_{2}\right)^{m} v_{2}^{\prime} \mapsto r p_{2} v_{3}
$$

By definition of sequential steps we have $m \leq n-1$. Because we have $\left.\left(f v_{3}\right)\right|_{r p_{3}}=\perp$, by induction hypothesis there exist $v^{\prime \prime \prime}, v^{\prime \prime \prime \prime} \in \llbracket \tau \rrbracket$ such that

$$
v^{\prime \prime} \mapsto_{r p_{2}} v^{\prime \prime \prime} \mapsto^{*} v^{\prime \prime \prime \prime}\left(\rightsquigarrow_{r p_{3}}\right)^{*} v_{3}
$$

That is, we get a sequence

$$
v_{1} \hookrightarrow r p_{1} v^{\prime} \mapsto^{*} v^{\prime \prime} \hookrightarrow r p_{2} v^{\prime \prime \prime} \hookrightarrow^{*} v^{\prime \prime \prime \prime}\left(\rightsquigarrow r p_{3}\right)^{*} v_{3}
$$

which is monotonic with respect to result positions.
If we have $v_{1}\left(\rightsquigarrow_{r p}\right)^{*} v_{2} \mapsto_{r p} v_{3}$ there exists $n \in \mathbb{N}_{0}$ such that $v_{1}\left(\rightsquigarrow_{r p_{1}}\right)^{n} v_{2} \mapsto_{r p_{1}} v_{3}$ with $\left.\left(f v_{3}\right)\right|_{r p_{3}}=\perp$.

Proof (of Lemma 4.3.5): As we consider only finite values we can prove the statement by induction over the number of constructors in $v$.

Base Case: For all $(\perp, r p)$ with $\left.(f \perp)\right|_{r p}=\perp$ we have $(\perp, r p) \in C_{f}$.
Inductive Case: Let $v \in \llbracket \tau \rrbracket$ and $r p \in \operatorname{Pos}(f v)$ such that $\left.(f v)\right|_{r p}=\perp$ and $(v, r p) \notin$ $C_{f}$. By Lemma 4.3.2 there exists $\left(v_{1}, r p^{\prime}\right)$ with $v_{1} \rightsquigarrow_{r p} v$ or $v_{1} \rightarrow_{r p^{\prime}} v$ and $r p^{\prime} \leq r p$. We distinguish two cases.
Case $1\left(v_{1} \rightsquigarrow r_{r p} v\right)$ : If $\left(v_{1}, r p\right) \in C_{f}$ then we set $v^{\prime}:=v_{1}$ and are finished. If $\left(v_{1}, r p\right) \notin C_{f}$ because $\left.\left(f v_{1}\right)\right|_{r p}=\perp$ by induction hypothesis there exists $\left(v_{2}, r p\right) \in C_{f}$ such that $v_{2}(\rightsquigarrow r p)^{*} v_{1}$ and, therefore, $v_{2}\left(\rightsquigarrow_{r p}\right)^{*} v$.
Case $2\left(v_{1} \mapsto_{r p^{\prime}} v\right)$ : Because $(v, r p) \notin C_{f}$ we have $\left(v_{1}, r p^{\prime}\right) \notin C_{f}$. Because $\left.\left(f v_{1}\right)\right|_{r p}=\perp$ by induction hypothesis there exists $\left(v_{2}, r p^{\prime}\right) \in C_{f}$ such that $v_{2}\left(\rightsquigarrow r p^{\prime}\right)^{*} v_{1}$. That is, we have a sequence of the form

$$
v_{2}\left(\rightsquigarrow_{r p^{\prime}}\right)^{*} v_{1} \mapsto_{r p^{\prime}} v
$$

Because $\left.(f v)\right|_{r p}=\perp$ by Lemma 4.3.4 there exist $v^{\prime}, v^{\prime \prime} \in \llbracket \tau \rrbracket$ such that

$$
v_{2} \mapsto_{r p^{\prime}} v^{\prime} \mapsto^{*} v^{\prime \prime}(\rightsquigarrow r p)^{*} v
$$

As this sequence is monotonic with respect to result positions $\left(v_{2}, r p\right) \in$ $C_{f}$ implies $\left(v^{\prime \prime}, r p\right) \in C_{f}$. That is, $\left(v^{\prime \prime}, r p\right) \in C_{f}$ and $v^{\prime \prime}(\rightsquigarrow r p)^{*} v$.

## A. 3 Proofs from Section 4.3.2

Proof (of Lemma 4.3.8): The functions $\bar{f}_{(v, r p)}$ and $f$ agree for total values because we have $\inf _{f} t v=f$ tv for all total values $t v$.
Let $v^{\prime} \in \llbracket \tau \rrbracket$. We show that we have $\bar{f}_{(v, r p)} v \sqsupset f v$ and $\bar{f}_{(v, r p)} v^{\prime} \sqsupseteq f v^{\prime}$ if $v^{\prime} \neq v$. We distinguish three cases.

Case $1\left(v^{\prime}=v\right)$ : Because $(v, r p) \in C_{f}$ we have $\left.(f v)\right|_{r p}=\perp$. Furthermore, because $\left.\left(\inf _{f} v\right)\right|_{r p} \sqsupset \perp$ we have

$$
\left.\left.\left.(f v)\right|_{r p} \sqcup\left(\inf _{f} v\right)\right|_{r p} \sqsupset(f v)\right|_{r p}
$$

By continuity of $\left.\cdot\right|_{r p}$ we get

$$
\left.\left.\left(f v \sqcup \inf _{f} v\right)\right|_{r p} \sqsupset(f v)\right|_{r p} .
$$

Because $(f v)[\cdot]_{r p}$ is strictly increasing and by a property of $\left.\cdot\right|_{r p}$ we get

$$
\bar{f}_{(v, r p)} v=(f v)\left[\left.\left(f v \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} \sqsupset(f v)\left[\left.(f v)\right|_{r p}\right]_{r p}=f v .
$$

Case $2\left(v^{\prime} \sqsupset v\right):$ We have

$$
f v^{\prime} \sqcup \inf _{f} v \sqsupseteq f v^{\prime}
$$

By monotonicity of $\left.\cdot\right|_{r p}$ we get

$$
\left.\left.\left(f v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p} \sqsupseteq\left(f v^{\prime}\right)\right|_{r p} .
$$

By monotonicity of $\left(f v^{\prime}\right)[\cdot]_{r p}$ we get

$$
\left(f v^{\prime}\right)\left[\left.\left(f v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} \sqsupseteq\left(f v^{\prime}\right)\left[\left.\left(f v^{\prime}\right)\right|_{r p}\right]_{r p} .
$$

Finally by a property of $\left.\cdot\right|_{r p}$ we get

$$
\left(f v^{\prime}\right)\left[\left.\left(f v^{\prime}\right)\right|_{r p}\right]_{r p}=f v^{\prime}
$$

and, therefore, we get

$$
\bar{f}_{(v, r p)} v^{\prime}=\left(f v^{\prime}\right)\left[\left.\left(f v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} \sqsupseteq f v^{\prime}
$$

Case $3\left(v^{\prime} \nsupseteq v\right)$ : By definition of $\bar{f}_{(v, r p)}$ we have $\bar{f}_{(v, r p)} v^{\prime}=f v^{\prime}$.
Lemma A.3.1: Let $v_{1} \in \llbracket \tau \rrbracket, v_{2} \in \llbracket \tau \rrbracket, v_{3} \in \llbracket \tau \rrbracket$, and $p \in \operatorname{Pos} v_{1}$ a position such that $v_{1} \lessdot \lessdot_{p} v_{2}$ and $v_{2} \sqsubseteq v_{3}$. For all $v \in \llbracket \tau \rrbracket$ with $v_{1} \sqsubseteq v, v \sqsubseteq v_{3}$, and $v_{2} \nsubseteq v$ we have $\left.v\right|_{p}=\perp$.
Proof: The following diagram illustrates the situation.


Additionally, $v_{1}$ and $v_{2}$ are related by $v_{1} \lessdot p v_{2}$ for some position $p$. We have $p \in \operatorname{Pos} v$ because $v_{1} \sqsubseteq v$ and $p \in \operatorname{Pos} v_{1}$. Because $v_{1} \lessdot_{p} v_{2}$, we have $\left.v_{1}\right|_{p}=\perp$ and there exists a constructor $C$ such that $\left.v_{2}\right|_{p}=C \perp$. As we furthermore have $v_{2} \sqsubseteq v_{3}$, the outermost constructor of $v_{3}$ at position $p$ is also $C$. Because $v \sqsubseteq v_{3}$, we have $\left.v\right|_{p}=\perp$ or the outermost constructor of $v$ at position $p$ is $C$. Finally, $v \sqsubseteq v_{3}$ and $v_{2} \nsubseteq v$ imply $\left.v\right|_{p}=\perp$.

Proof (of Lemma 4.3.9): If $(v, r p) \in C_{f}$ there exists a sequence of the form $\perp \mapsto^{n} v$. By induction on $n$ we show that $v$ is finite.

Base Case $\left(\perp \mapsto^{0} \perp\right)$ : Let $\left\langle v_{i}\right\rangle_{i \in I}$ be a chain such that $\perp \sqsubseteq \bigsqcup_{i \in I} v_{i}$. We have $\perp \sqsubseteq v_{i}$ for all $i \in I$. Thus, $\perp$ is finite.

Inductive Case $\left(\perp \mapsto^{n+1} v\right)$ : Let $\left\langle v_{i}\right\rangle_{i \in I}$ be a chain with $v \sqsubseteq \bigsqcup_{i \in I} v_{i}$. There exists $v^{\prime}$ such that $\perp \mapsto^{n} v^{\prime} \hookrightarrow v$. Because $v^{\prime} \sqsubseteq v$ we have $v^{\prime} \sqsubseteq \bigsqcup_{i \in I} v_{i}$. By induction hypothesis there exists $i \in I$ such that $\overline{v^{\prime}} \sqsubseteq v_{i}$. We consider a subset $J$ of $I$ such that $J=\left\{i \in I \mid v_{i} \sqsupseteq v^{\prime}\right\}$. As there exists at least one element $v_{i} \sqsubseteq v^{\prime}$ we have $\bigsqcup_{i \in I} v_{i}=\bigsqcup_{j \in J} v_{j}$.
We assume that we have $v \nsubseteq v_{j}$ for all $j \in J$. Let $j \in J$. We have $v^{\prime} \lessdot_{p} v, v^{\prime} \sqsubseteq v_{j}$, $v \nsubseteq v_{j}$, and $v_{j} \sqsubseteq \bigsqcup_{j \in J} v_{j}$ and $v \sqsubseteq \bigsqcup_{j \in J} v_{j}$. The following diagram illustrates the situation.


By Lemma A.3.1 we get $\left.v_{j}\right|_{p}=\perp$. That is, we have $\left.v_{j}\right|_{p}=\perp$ for all $j \in J$. Because $\left.\cdot\right|_{p}$ is continuous we get

$$
\left.\left(\bigsqcup_{j \in J} v_{j}\right)\right|_{p}=\bigsqcup_{j \in J}\left(\left.v_{j}\right|_{p}\right)=\perp
$$

This is a contradiction to $v \sqsubseteq \bigsqcup_{i \in I} v_{i}$ as we have $\left.v\right|_{p} \neq \perp$. Thus there exists $j \in J$ with $v \sqsubseteq v_{j}$, that is, $v$ is finite.

Proof (of Lemma 4.3.10): Let $v_{1}, v_{2} \in \llbracket \tau \rrbracket$ with $v_{1} \sqsubseteq v_{2}$. We distinguish two cases.
Case $1\left(v_{1} \nexists v\right)$ : We reason as follows and employ monotonicity of $f$ and $\bar{f} \prec f$.

$$
\bar{f} v_{1}=f v_{1} \sqsubseteq f v_{2} \sqsubseteq \bar{f} v_{2}
$$

Case $2\left(v_{1} \sqsupseteq v\right)$ : Because $v_{1} \sqsupseteq v$ and $v_{1} \sqsubseteq v_{2}$ we have $v_{2} \sqsupseteq v$. By $v_{1} \sqsubseteq v_{2}$ and monotonicity of $f$ we get

$$
f v_{1} \sqsubseteq f v_{2} .
$$

By monotonicity of $\cdot \sqcup \inf _{f} v$ we get

$$
f v_{1} \sqcup \inf _{f} v \sqsubseteq f v_{2} \sqcup \inf _{f} v
$$

and by monotonicity $\left.\cdot\right|_{r p}$ we have

$$
\left.\left.\left(f v_{1} \sqcup \inf _{f} v\right)\right|_{r p} \sqsubseteq\left(f v_{2} \sqcup \inf _{f} v\right)\right|_{r p} .
$$

By monotonicity of $\left(f v_{1}\right)[\cdot]_{p}$ we get

$$
\left(f v_{1}\right)\left[\left.\left(f v_{1} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} \sqsubseteq\left(f v_{1}\right)\left[\left.\left(f v_{2} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p}
$$

and by monotonicity of $\cdot\left[\left.\left(f v_{2} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{p}$ we get

$$
\left(f v_{1}\right)\left[\left.\left(f v_{2} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} \sqsubseteq\left(f v_{2}\right)\left[\left.\left(f v_{2} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} .
$$

This final inequality shows that we have $\bar{f} v_{1} \sqsubseteq \bar{f} v_{2}$.
Proof (of Lemma 4.3.11): Let $\left\langle v_{i}\right\rangle_{i \in I}$ be a chain in $\llbracket \tau \rrbracket$. We have to prove the following equality.

$$
\bar{f}_{(v, r p)}\left(\bigsqcup_{i \in I} v_{i}\right)=\bigsqcup_{i \in I}\left(\bar{f}_{(v, r p)} v_{i}\right)
$$

Case $1\left(\bigsqcup_{i \in I} v_{i} \nsupseteq v\right)$ : In this case we have $v_{i} \nsupseteq v$ for all $i \in I$ and reason as follows.

$$
\begin{aligned}
\bar{f}\left(\bigsqcup_{i \in I} v_{i}\right) & =f\left(\bigsqcup_{i \in I} v_{i}\right) & & \text { definition of } \bar{f} \\
& =\bigsqcup_{i \in I}\left(f v_{i}\right) & & \text { continuity of } f \\
& =\bigsqcup_{i \in I}\left(\bar{f} v_{i}\right) & & \text { definition of } \bar{f}, v_{i} \nexists v
\end{aligned}
$$

Case $2\left(\bigsqcup_{i \in I} v_{i} \sqsupseteq v\right)$ : We consider a subset $J$ of $I$ such that $J=\left\{i \in I \mid v_{i} \sqsupseteq v\right\}$. By Lemma 4.3.9 $v$ is finite. That is, there exists $i \in I$ such that $v_{i} \sqsupseteq v$ and, therefore, $J$ is not empty. Because $\left\langle v_{i}\right\rangle_{i \in I}$ is a chain we only remove dominated elements, that is, $\bigsqcup_{i \in I} v_{i}=\bigsqcup_{i \in J} v_{i}$. Furthermore, by monotonicity of $\bar{f}$ we have $\bigsqcup_{i \in I}\left(\bar{f} v_{i}\right)=\bigsqcup_{i \in J}\left(\bar{f} v_{i}\right)$. We reason as follows.

$$
\begin{aligned}
\bar{f}\left(\bigsqcup_{i \in I} v_{i}\right) & =\bar{f}\left(\bigsqcup_{i \in J} v_{i}\right) & & \text { supremum property } \\
& =\left(f\left(\bigsqcup_{i \in J} v_{i}\right)\right)\left[\left.\left(f\left(\bigsqcup_{i \in J} v_{i}\right) \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} & & \text { definition of } \bar{f} \\
& \left.=\left.\left(\bigsqcup_{i \in J} f v_{i}\right)\left[\left(\bigsqcup_{i \in J} f v_{i}\right) \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} & & \text { continuity of } f \\
& =\left(\bigsqcup_{i \in J} f v_{i}\right)\left[\left.\left(\bigsqcup_{i \in J}\left(f v_{i} \sqcup \inf _{f} v\right)\right)\right|_{r p}\right]_{r p} & & \text { supremum property } \\
& =\left(\bigsqcup_{i \in J} f v_{i}\right)\left[\left.\bigsqcup_{i \in J}\left(f v_{i} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p} & & \text { continuity of }\left.\cdot\right|_{r p} \\
& =\bigsqcup_{i \in J}\left(f v_{i}\right)\left[\left(\left.f v_{i} \sqcup \inf _{f} v\right|_{r p}\right]_{r p}\right) & & \text { continuity of } \cdot[\cdot]_{r p}
\end{aligned}
$$

$$
\begin{aligned}
& =\bigsqcup_{i \in J}\left(\bar{f} v_{i}\right) \\
& =\bigsqcup_{i \in I}\left(\bar{f} v_{i}\right)
\end{aligned}
$$

definition of $\bar{f}, v_{i} \sqsupseteq v$
supremum property

This shows that $\bar{f}$ is continuous.
Proof (of Lemma 4.3.12): We only have to consider the case that there exists $p v^{\prime} \in$ $\llbracket \tau \rrbracket$ with $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p}=\perp$ and $p v^{\prime} \sqsupseteq v$. Otherwise $p$ already satisfies the requirements we are looking for and we can set $p^{\prime}:=p$.

Let $(v, r p) \in C_{\tau}$ with $v \nsubseteq p v$. Then there exists a sequence of sequential steps of the form $\perp \hookrightarrow^{*} v$. There exists a sub-sequence of the form $v_{1} \mapsto^{*} v_{2}$ with $v_{1} \sqsubseteq p v$ and $v_{2} \nsubseteq p v$ because $\perp \sqsubseteq p v$ and $v \nsubseteq p v$. We consider such a sequence of minimal length. This sequence has length one because there exists no value $v^{\prime}$ such that $v^{\prime} \nsubseteq p v$ and $v^{\prime} \sqsubseteq p v$. Therefore, we have a step of the form $v_{1} \rightarrow_{r p^{\prime}} v_{2}$ with $v_{1} \sqsubseteq p v$ and $v_{2} \nsubseteq p v$. The following Hasse-diagram illustrates the situation.


Let $p^{\prime}$ be the corresponding sequential position of $v_{1} \hookrightarrow_{r p^{\prime}} v_{2}$. We show that $p^{\prime}$ is a sequential position in $p v$ at position $r p$. We have $v_{1} \lessdot p^{\prime} v_{2}, v_{2} \sqsubseteq p v^{\prime}, v_{1} \sqsubseteq p v$, $p v \sqsubseteq p v^{\prime}$, and $v_{2} \nsubseteq p v$. By Lemma A.3.1 we have $\left.p v\right|_{p^{\prime}}=\perp$.

Because $p^{\prime}$ is a sequential position in $v_{1}$ at position $r p^{\prime}$ and we have $v_{1} \sqsubseteq p v$ and $\left.p v\right|_{p^{\prime}}=\perp$ by definition of sequential positions $p^{\prime}$ is also a sequential position in $p v$ at position $r p^{\prime}$. By definition of the characteristic set $C_{f}$ we have $r p \geq r p^{\prime}$ which implies $r p=r p^{\prime}$ because $\left.(f p v)\right|_{r p}=\perp$ and $\left.(f p v)\right|_{r p^{\prime}}=\perp$.

That is, $p^{\prime}$ is a sequential position in $p v$ at position $r p$. Because $\left.v_{2}\right|_{p^{\prime}} \neq \perp$ and $v \sqsupseteq v_{2}$ we have $\left.v\right|_{p^{\prime}} \neq \perp$. This implies $p v^{\prime} \not \equiv v$ for all $p v^{\prime} \in \llbracket \tau \rrbracket$ with $\left.p v^{\prime}\right|_{p^{\prime}}=\perp$.
Proof (of Lemma 4.3.13): Let $p v$ be a partial value of type $\tau$ and $r p^{\prime} \in \operatorname{Pos}(\bar{f} p v)$ such that $\left.(\bar{f} p v)\right|_{r p^{\prime}}=\perp$. We have to show that there exists $p \in \operatorname{Pos} p v$ such that $\left.p v\right|_{p}=\perp$ and for all $p v^{\prime} \in \llbracket \tau \rrbracket$ the following holds.

$$
\left.p v^{\prime} \sqsupseteq p v \wedge p v^{\prime}\right|_{p}=\left.\perp \Longrightarrow\left(\bar{f} p v^{\prime}\right)\right|_{r p^{\prime}}=\perp
$$

We distinguish two cases.

Case $1(p v \sqsupseteq v)$ : We have

$$
\perp=\left.(\bar{f} p v)\right|_{r p^{\prime}}=\left.\left((f p v)\left[\left.\left(f p v \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p}\right)\right|_{r p^{\prime}}
$$

and distinguish further sub-cases.
Case a $\left(r p^{\prime} \geq r p\right)$ : We reason as follows.

$$
\begin{array}{rlrl}
\left.\left((f p v)\left[\left.\left(f p v \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p}\right)\right|_{r p^{\prime}} & & \\
& =\left.\left(f p v \sqcup \inf _{f} v\right)\right|_{r p^{\prime}} & & r p^{\prime} \geq r p \\
& =\left.\left.(f v)\right|_{r p^{\prime}} \sqcup\left(\inf _{f} v\right)\right|_{r p^{\prime}} & & \text { continuity of }\left.\cdot\right|_{r p^{\prime}}
\end{array}
$$

This implies $\left.(f p v)\right|_{r p^{\prime}}=\perp$ and $\left.\left(\inf _{f} v\right)\right|_{r p^{\prime}}=\perp$. Because $f$ is sequential, there exists a sequential position $p$ such that $\left.p v\right|_{p}=\perp$.

Let $p v^{\prime} \in \llbracket \tau \rrbracket$ with $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p}=\perp$. The following equality proves that $p$ is a sequential position in this case.

$$
\begin{aligned}
\left.\left(\bar{f} p v^{\prime}\right)\right|_{r p^{\prime}} & =\left.\left(\left(f p v^{\prime}\right)\left[\left.\left(f p v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p}\right)\right|_{r p^{\prime}} & & \text { definition of } \bar{f}, p v^{\prime} \sqsupseteq p v \\
& =\left.\left(f p v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p^{\prime}} & & r p^{\prime} \geq r p \\
& =\left.\left.\left(f p v^{\prime}\right)\right|_{r p^{\prime}} \sqcup\left(\inf _{f} v\right)\right|_{r p^{\prime}} & & \text { continuity of }\left.\cdot\right|_{r p^{\prime}} \\
& =\left.\perp \sqcup\left(\inf _{f} v\right)\right|_{r p^{\prime}} & & p \text { sequential position } \\
& =\perp & & \left.\left(\inf _{f} v\right)\right|_{r p^{\prime}}=\perp
\end{aligned}
$$

Case b(r $\left.p^{\prime} \nsupseteq r p \wedge r p \nsupseteq r p^{\prime}\right)$ : We have

$$
\left.\left((f p v)\left[\left.\left(f p v \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p}\right)\right|_{r p^{\prime}}=\left.(f p v)\right|_{r p^{\prime}}
$$

Because $f$ is a sequential function, there exists a sequential position $p$ such that $\left.p v\right|_{p}=\perp$.

Let $p v^{\prime} \in \llbracket \tau \rrbracket$ with $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p}=\perp$. We reason as follows.

$$
\begin{aligned}
\left.\left(\bar{f} p v^{\prime}\right)\right|_{r p^{\prime}} & =\left.\left(\left(f p v^{\prime}\right)\left[\left.\left(f p v^{\prime} \sqcup \inf _{f} v\right)\right|_{r p}\right]_{r p}\right)\right|_{r p^{\prime}} & & \text { definition of } \bar{f}, p v^{\prime} \sqsupseteq p v \\
& =\left.\left(f p v^{\prime}\right)\right|_{r p^{\prime}} & & r p^{\prime} \nsupseteq r p \wedge r p \nsupseteq r p^{\prime} \\
& =\perp & & p \text { sequential position }
\end{aligned}
$$

Case $2(p v \nsupseteq v)$ : We have

$$
\perp=\left.(\bar{f} p v)\right|_{r p}=\left.(f p v)\right|_{r p}
$$

Because $f$ is sequential, there exists a sequential position $p$ in $p v$ at position $r p$. By Lemma 4.3.12 there exists a sequential position $p^{\prime}$ in $p v$ at position $r p$ such that for all $p v^{\prime} \in \llbracket \tau \rrbracket$ with $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p^{\prime}}=\perp$ we have $p v^{\prime} \nsupseteq v$.

Let $p v^{\prime}$ such that $p v^{\prime} \sqsupseteq p v$ and $\left.p v^{\prime}\right|_{p^{\prime}}=\perp$. We show that $p^{\prime}$ is a sequential position as follows.

$$
\begin{aligned}
\left.\left(\bar{f} p v^{\prime}\right)\right|_{r p} & =\left.\left(f p v^{\prime}\right)\right|_{r p} & & p v^{\prime} \nexists v \\
& =\perp & & p^{\prime} \text { sequential position }
\end{aligned}
$$

This completes the proof that $\bar{f}$ is a sequential function.

A Proofs from Chapter 4

## B Proofs from Chapter 6

## B.1 Proofs from Section 6.3

To prove Lemma 6.3.1 we define a function strictConst that behaves like the constant function const but which is strict in its second argument.
Definition B.1.1 (StrictConst): The function strictConst :: $\alpha \rightarrow \beta \rightarrow \alpha$ satisfies

$$
\text { strictConst } x y=\left\{\begin{array}{ll}
\perp & y=\perp \\
x & \text { otherwise }
\end{array}\right. \text {. }
$$

We can define strictConst as strictConst $x y=\operatorname{seq} y x$. If we are not equipped with seq we can as well define a function strictConst $\tau:: \alpha \rightarrow \tau \rightarrow \alpha$ for every type $\tau$ by pattern matching. The functions strictConst and ${\text { strict } \text { Const }_{\tau} \text { are strict and total with }}$ respect to to their second argument.
Proof (of Lemma 6.3.1): Let $x_{s}::[()]$. Let $y$ be an element of type $\tau$. We consider the list $y s=$ map (strictConst $y) x$ s. That is, we replace all occurrences of () in $x$ sy $y$ and all occurrences of $\perp_{()}$by $\perp_{\tau}$. We can reconstruct $x s$ from $y s$ if we replace all occurrences of $y$ in $y s$ by () and all occurrences of $\perp_{\tau}$ by $\perp_{()}$, that is, $x s \equiv$ map (strictConst ()) ys. We reason as follows.

```
\(f_{()} x s\)
    \(\equiv\{x s \equiv\) map (strictConst ()) ys \(\}\)
\(f_{()}(\)map (strictConst ()) ys)
    \(\equiv\{\) free theorem for \(f\), strictConst () strict \(\}\)
map (strictConst ()) ( \(f_{\tau}\) ys)
    \(\sqsupseteq \quad\left\{\right.\) monotonicity and \(\left.f_{\tau} \preceq g_{\tau}\right\}\)
    map (strictConst ()) (g ys)
    \(\equiv\) \{ free theorem for \(g\), strictConst () strict \}
\(g_{()}(\)map (strictConst ()) ys)
    \(\equiv\{x s \equiv\) map (strictConst ()) ys \(\}\)
    \(g_{()} x\) s
```

Thus, we have $f_{()} x s \sqsupseteq g_{()} x s$ for all $x s::[()]$, that is, $f_{()} \preceq g_{()}$.
Definition B.1.2 (Free Theorem Basics): In this definition we present some of the basic definitions by Johann and Voigtländer (2004).
strict A relation $R$ is strict if $(\perp, \perp) \in R$.
bottom-reflecting A relation $R$ is bottom-reflecting if for every $(x, y) \in R$ we have $x \neq \perp$ if and only if $y \neq \perp$.
continuous A relation $R$ is continuous if for all chains $\left\langle x_{i}\right\rangle_{i \in I}$ and $\left\langle y_{i}\right\rangle_{i \in I}$ whose elements are pair-wise related by $R$ we have $\left(\bigsqcup_{i \in I} x_{i}, \bigsqcup_{i \in I} y_{i}\right) \in R$.

To prove Lemma 6.3.2 we prove the following generalization of the functional free theorem for a function of type $f::(\alpha \rightarrow \beta) \rightarrow[\alpha] \rightarrow[\beta]$. The standard functional free theorem states that

$$
h(p x) \equiv q(g x)
$$

for all $x:: \tau_{1}$ implies

$$
\operatorname{maph}(f p x s) \equiv f q(\operatorname{map} g x s)
$$

for all $x s::\left[\tau_{1}\right]$. In contrast, the following lemma shows that we do not have to show $h(p x) \equiv q(g x)$ for all $x:: \tau_{1}$ but only for the elements of $x$.

Lemma B.1.1: Let $f::(\alpha \rightarrow \beta) \rightarrow[\alpha] \rightarrow[\beta]$. For strict functions $g:: \tau_{1} \rightarrow \tau_{2}, h:: \tau_{3} \rightarrow$ $\tau_{4}, p:: \tau_{1} \rightarrow \tau_{3}$, and $q:: \tau_{2} \rightarrow \tau_{4}$ such that

$$
h(p(x s!!n)) \equiv q(g(x s!!n))
$$

for all $n$ :: Int we have

$$
\operatorname{map} h(f p x s) \equiv f q(\operatorname{map} g x s)
$$

for all $x s::\left[\tau_{1}\right]$.
Proof: Let $x s::\left[\tau_{1}\right]$. We define the relations

$$
R_{x s}:=\left\{\left(\bigsqcup_{i \in I}(x s!!i), \bigsqcup_{i \in I} g(x s!!i)\right) \mid\langle x s!!i\rangle_{i \in I} \text { chain, } I \subseteq I n t\right\}
$$

and $S:=\{(x, h x)\}$. Theorem 6.4.1 shows that $R_{x s}$ is strict and continuous. Furthermore, $S$ is strict and continuous because $h$ is strict and continuous.

The relational free theorem for $f::(\alpha \rightarrow \beta) \rightarrow[\alpha] \rightarrow[\beta]$ states that, if for all $(x, y) \in R_{x s}$ we have $(p x, q y) \in S$, then for all $(x s, y s) \in \operatorname{lift}\{[]\}\left(R_{x s}\right)$ we have $(f p x s, f q y s) \in \operatorname{lift}\{[]\}(S)$. Let $(x, y) \in R_{x s}$. To show that $(p x, q y) \in S$ we have to show that $h(p x) \equiv q y$. By definition of $R_{x s}$ we have $x=\bigsqcup_{i \in I}(x s!!i)$ and $y=\bigsqcup_{i \in I} g(x s!!i)$. We reason as follows.

$$
\begin{aligned}
h(p(\bigsqcup x s!!i)) & \equiv \bigsqcup h(p(x s!!i)) & & \text { continuity } \\
& \equiv \bigsqcup q(g(x s!!i)) & & \text { precondition } \\
& \equiv q(\bigsqcup g(x s!!i)) & & \text { continuity }
\end{aligned}
$$

That is, $h(p x) \equiv q y$.
Thus, for all $(x s, y s) \in \operatorname{lift}\{[]\}\left(R_{x s}\right)$ we as well have $(f p x s, f q y s) \in \operatorname{lift}\{[]\}(S)$. As we have $(x s$, map $g x s) \in \operatorname{lift}\{[]\}\left(R_{x s}\right)$ we get $(f p x s, f q($ map $g x s)) \in \operatorname{lift}\{[]\}(S)$, that is, maph $(f p x s) \equiv f q(\operatorname{map} g x s)$.

The following lemma shows a simple property about $[n .$.$] and (!!). Here and in$ the following we consider $[n .$.$] as shortform of iterate (+1) n$.

Lemma B.1.2: For all $n::$ Int and all $m::$ Int with $m \geq 0$ we have $[n .]!!.m \equiv n+m$.
Proof: By induction over $m$.
Base Case:

```
iterate (+1) n !! 0
    { {definition of iterate }
(n:iterate (+1) (n+1))!!0
    \equiv{definition of (!!)}
n
```


## Inductive Case:

```
iterate \((+1) n!!(m+1)\)
    \(\equiv\) \{definition of iterate \(\}\)
\((n\) : iterate \((+1)(n+1))!!(m+1)\)
\(\equiv\{\) definition of \((!!)\}\)
iterate \((+1)(n+1)!!m\)
    \(\equiv\) \{induction hypothesis \}
\(n+1+m\)
```

Proof (of Lemma 6.3.2): Let $x:: \tau$ and $x s::[\tau]$. We reason as follows.

```
map ((x:xs)!!) [1..]
    \equiv{definition of [1..]}
map ((x:xs)!!) (iterate (+1) 1)
    \equiv {free theorem for iterate, (+1) strict & total }
map ((x:xs)!!) (map (+1) (iterate (+1) 0))
    \equiv {Lemma B.1.1 and Lemma B.1.2 }
map (xs!!) (iterate (+1) 0)
    \equiv{definition of [0..]}
map (xs!!) [0..]
```

To prove the step labeled with $(*)$ we have to prove the following equality.

$$
(x: x s)!!([1 . .]!!n) \equiv x s!!([0 . .]!!n)
$$

To prove this statement we distinguish two cases.
Case $1(n \equiv \perp \vee n<0)$ : We reason as follows.

```
(x:xs)!! ([1..]!!n)
\equiv{definition of (!!)}
(x:xs)!! \perp
    \equiv{definition of (!!)}
xs !! \perp
    \equiv{definition of (!!)}
xs !! ([0..]!!n)
```

Case $2(n \geq 0)$ : We reason as follows.

$$
\begin{aligned}
& (x: x s)!!([1 . .]!!n) \\
& \equiv\{\text { Lemma B.1.2 }\} \\
& (x: x s)!!n+1 \\
& \equiv\{\text { definition of }(!!)\} \\
& x s!!n \\
& \equiv\{\text { Lemma B.1.2 }\} \\
& x s!!([0 \ldots]!!n)
\end{aligned}
$$

This step finally concludes the proof.
Proof (of Lemma 6.3.6): Let $x s::[\tau]$ and $n::$ Int such that $f_{\tau} x s!!n \not \equiv \perp, g_{\tau} x s!!n \not \equiv$ $\perp$, and $f_{\tau} x s!!n \not \equiv g_{\tau} x s$ !! $n$. By Lemma 6.3 .5 there exist $k, l::$ Int such that the following holds.

$$
x s!!k \equiv f_{\tau} x s!!n \not \equiv g_{\tau} x s!!n \equiv x s!!l
$$

This implies $k \not \equiv l$. Furthermore we have $k \not \equiv \perp$ and $l \not \equiv \perp$ because $f_{\tau} x s!!n \not \equiv \perp$ and $g_{\tau} x s!!n \not \equiv \perp$. This also implies $x s!!k \not \equiv \perp$ and $x s!!l \not \equiv \perp$.

If we replace the element of $x s$ at position $k$ by $\perp$ we get a list $y s::[\tau]$ with $y s$ !! $k \equiv \perp$ and $y s!!l \not \equiv \perp$. We have shape $x s \equiv$ shape $y s$ and by Lemma 6.3 .5 we get $f_{\tau} y s!!n \equiv y s!!k \equiv \perp$ and $g_{\tau}$ ys $!!n \equiv y s!!l \equiv x s!!l \not \equiv \perp$.

Proof (of Lemma 6.3.7): Because we have $f_{\tau} \npreceq g_{\tau}$ there exists $x s::[\tau]$ such that $f_{\tau} x s \not \equiv g_{\tau} x$ s. We distinguish three cases.

Case 1 (shape $\left(f_{\tau} x s\right) \nexists$ shape $\left(g_{\tau} x s\right)$ ): We set $y s \equiv$ map (strictConst ()$) x s$ and reason as follows.

```
shape (f) ys)
    \equiv{ys \equivmap (strictConst ()) xs }
shape (f() (map (strictConst ()) xs))
    \equiv {free theorem for f, strictConst () strict }
shape (map (strictConst ()) (f\tauxs))
    \equiv{free theorem for shape, strictConst () strict }
shape ( }\mp@subsup{f}{\tau}{}xs
#
shape (g\tau xs)
    \equiv{ free theorem for shape, strictConst () strict }
shape (map (strictConst ()) (g\tau xs))
    \equiv {free theorem for g, strictConst () strict }
shape (g() (map (strictConst ()) xs))
    \equiv{ys \equivmap (strictConst ()) xs }
shape (g() ys)
```

Therefore, there exists ys :: [()] such that shape $\left(f_{()} y s\right) \not \equiv$ shape $\left(g_{()} y s\right)$ which implies $f_{()} \npreceq g_{()}$.

Case 2 ( $\exists n::$ Int. $\left.f_{\tau} x s!!n \equiv \perp \wedge g_{\tau} x s!!n \not \equiv \perp\right):$ We set $y s \equiv \operatorname{map}$ (strictConst ()) xs and reason as follows.

```
f() ys !!n
    \equiv {ys \equivmap (strictConst ()) xs }
f() (map (strictConst ()) xs)!!n
    \equiv{free theorem for f, strictConst () strict }
    map (strictConst ()) (f\tau xs) !! n
    \equiv { free theorem for (!!), strictConst () strict }
strictConst () ( }\mp@subsup{f}{\tau}{}xs!!n
    \equiv{definition of strictConst, f\tau xs !! n\equiv\perp}
\perp
```

For $g$ we reason the same way.

```
g() ys !!n
    \equiv {ys \equivmap (strictConst ()) xs }
g()
    \equiv {free theorem for g, strictConst () strict }
    map (strictConst ()) (g\tau xs)!!n
    \equiv{ free theorem for (!!), strictConst () strict }
strictConst () (g\tau xs !! n)
    #\equiv { definition of strictConst, g\tau xs !! n #三 \perp}
    \perp
```

That is, there exists $y s::[()]$ and $n::$ Int such that $f_{()}$ys $!!n \equiv \perp$ and $g_{()}$ys $!!n \not \equiv$ $\perp$ which implies $f_{()} \npreceq g_{()}$.

Case 3 ( $\exists n::$ Int. $\left.f_{\tau} x s!!n \not \equiv \perp \wedge g_{\tau} x s!!n \not \equiv \perp \wedge f_{\tau} x s!!n \not \equiv g_{\tau} x s!!n\right)$ : According to Lemma 6.3.6 there exists $y s::[\tau]$ and $n::$ Int such that $f_{\tau}$ ys $!!n \equiv \perp$ and $g_{\tau}$ ys !! $n \not \equiv \perp$. That is, we reason as in Case 2.

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[^0]:    ${ }^{1}$ An extended version is available as technical report (Chitil 2011).

[^1]:    ${ }^{1}$ Here and in the following, we use the symbol $\rightarrow$ for the token $->$.

[^2]:    ${ }^{2}$ In Haskell programs the symbol $\lambda$ is, in fact, represented by a backslash.

[^3]:    ${ }^{3}$ In Haskell we can even choose the implementation depending on the result type of the function.
    ${ }^{4}$ In fact, $E q$ also provides an inequality check $(/=):: \alpha \rightarrow \alpha \rightarrow$ Bool that is left out for simplicity.

[^4]:    ${ }^{1}$ Hackage is a database for Haskell libraries and is available at http://hackage. haskell. org.
    ${ }^{2}$ The word "mathematics" occurs three times in Shakespeare's collected works.

[^5]:    ${ }^{3}$ For readibility we use $\perp$ in the output while Sloth actually uses the symbol _ instead.

[^6]:    ${ }^{1}$ Accidentally Chitil (2006) uses $\bigsqcup$ instead of $\sqcap$.

[^7]:    5 Implementation of Sloth

[^8]:    ${ }^{1}$ The function seq $:: \alpha \rightarrow \beta \rightarrow \beta$ satisfies the laws seq $\perp y \equiv \perp$ and $\operatorname{seq} x y \equiv y$ if $x \not \equiv \perp$.

[^9]:    ${ }^{2}$ The function const $:: \alpha \rightarrow \beta \rightarrow \alpha$ is defined by const $x_{-}=x$.

[^10]:    ${ }^{3}$ A function $f$ is strict if $f \perp \equiv \perp$.

[^11]:    ${ }^{4}$ A function $f$ is total if for all $x$ with $x \not \equiv \perp$ we have $f x \not \equiv \perp$.

[^12]:    ${ }^{5}$ A function $f$ is total if for all $x$ with $x \not \equiv \perp$ we have $f x \not \equiv \perp$.

[^13]:    ${ }^{6}$ The generator is available at www-ps.iai.uni-bonn.de/cgi-bin/polyseq.cgi.

[^14]:    ${ }^{7}$ Here strict means $f x \perp \equiv \perp$ and $f \perp y \equiv \perp$ for all $x, y$.

[^15]:    ${ }^{1}$ The type Int uses 32 or 64 bits depending on the architecture of the system, the GHC is running on.

[^16]:    ${ }^{2}$ The run-times are measured with the Curry compiler KiCS (Braßel and Huch 2009).

[^17]:    ${ }^{1}$ Rules can be considered as syntactic sugar for case expressions.

[^18]:    ${ }^{2} \mathrm{~A}$ corrected version is available as technical report (Christiansen et al. 2011b).

