

Chapter 1

Preface

1.1 Motivation

The halting problem is undecidable in general, however this property is often abused to deduce that for all programs. The intent of this project is to explore some context in which the halting property *is* decidable, and to analyze how useful this indeed is.

1.2 Expectations of the reader

The reader is expected to have a background in computer science on a graduate level or higher. In particular, it is expected that the reader is familiar with basic concepts of compilers, computability and complexity, which are subject to basic undergraduate courses at the state of writing. Furthermore, the reader is expected to be familiar with discrete mathematics and the concepts of functional programming languages. Ideally, the reader should know at least one purely functional programming language.

For those still in doubt, it is expected that the following terms can be used without definition:

- Algorithm, Recursion, Induction, Big O Notation
- Regular Expressions (preg syntax)
- Backus-Naur Form
- Turing Machine, Halting Problem
- List, Binary tree, Head, Tail
- Discrete mathematics

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Chapter 2

On the general uncomputability of the halting problem

2.1 Computable problems and effective procedures

A computable problem is a problem that can be solved by an effective procedure.

A problem can be solved by an effective procedure iff the effective procedure is well-defined for the entire problem domain¹, and iff passing a value from the domain as input to the procedure *eventually* yields a correct result (to the problem) as output of the procedure. That is, an effective procedure can solve a problem if it computes an injective partial function that associates the problem domain with the range of solutions to the problem.

An effective procedure is discrete, in the sense that computing the said function cannot take an infinite amount of time. To do this, an effective procedure makes use of a finite sequence of steps that themselves are discrete. This has a few inevitable consequences for the input and output values, namely that they themselves must be discrete and that there must be a discrete number of them².

Proof. An infinite value cannot be processed nor produced by a finite sequence of discrete steps. □

An effective procedure is also deterministic, in the sense that passing the same input value always yields the same output value. This means that all of the steps of the procedure that are relevant to it's output³ are themselves deterministic.

Proof. If a procedure made use of a stochastic process to yield a result, that stochastic process would have to yield the output for the same input if the global deterministic property of the procedure is to be withheld. This is clearly absurd. □

In effect, a procedure can be said to comprise of a finite sequence of other procedures, which themselves may comprise of other procedures, however, all procedures eventually bottom out, in that a finite sequence of composite procedures can always be replaced by a finite sequence of basic procedures that are implemented in underlying hardware.

- effective procedure
- effectively decidable
- effectively enumerable

¹Invalid inputs are, in this instance, irrelevant.

²A finite sequence of discrete values can be trivially encoded as a single discrete value.

³All other steps can be omitted without loss of generality.

2.2 Enumerability

2.2.1 Enumerable sets

Enumerable sets, or equivalently countable or recursively enumerable sets, are sets that can be put into a one-to-one correspondence to the set of natural numbers \mathbb{N} , more specifically:

Definition 1. *An enumerable set is either the empty set or a set whose elements can be placed in a sequence s.t. each element gets a consecutive number from the set of natural numbers \mathbb{N} .*

2.2.2 Decidability

Definition 2. *A problem is decidable if there exists an algorithm that for any input event*

- Recursively enumerable – countable sets
- Co-recursively enumerable

2.3 Cantor's diagonalization

Cantor's diagonalization argument is a useful argument for proving unenumerability of a set and hence its uncomputability.

The original proof shows that the set of infinite bit-sequences is not enumerable.

Proof. Assume that sequence S is an infinite sequence of infinite sequences of bits. The claim is that regardless of the number of bit-sequences in S it is always possible to construct a bit-sequence not contained in S .

Such a sequence can be represented as a table:

Such a sequence is constructable by taking the complements of the elements along the diagonal of all

□

2.4 The halting problem

2.5 Rice's statement

2.6 Primitive recursion

All primitive recursive programs terminate.

2.7 Introduction to size-change termination

The size change termination .. why values should be well-founded

2.8 The language to be defined

The soft version.

Chapter 3

D

For the purposes of describing size-change termination we'll consider a language D. The following chapter describes the syntax and semantics of the language.

3.1 General properties

The intent of the language is for it be used to explain concepts such as size-change termination. One of the fundamental concepts required of the language of application is that it's datatypes are well-founded. That is, any subset S of the range of values of some well-defined type has a value s s.t. $\forall s' \in S \ s \leq s'$. This makes it ideal to chose some oversimplistic data type structure rather than an army of basic types. Besides, an appropriately defined basic data type should be able to represent arbitrarily complex data values.

The language is initially first-order since the size-change termination principle is first described for first-order programs later on in this work. However, the language is designed so that it is easy to turn it into a high-level language without much effort. This may prove necessary as we try to expand size-change termination to higher-order programs.

The language is a call-by-value and purely functional to avoid any problems that could arise from regarding lazy programs or where the notion of a global state of the machine is relevant. Simply put, this is done to ensure elegance of further proof with the help of the language.

3.2 Syntax

We describe the syntax of D in terms of an extended Backus-Naur form¹. This is a core syntax definition, and other, more practical, syntactical features may be defined later on as needed.

$$\langle \text{expression} \rangle ::= \langle \text{value} \rangle (\text{'.'} \langle \text{expression} \rangle) ? \quad (3.1)$$

$$\langle \text{value} \rangle ::= \text{'0'} \mid \text{'('} \langle \text{value} \rangle \text{'})' \mid \langle \text{application} \rangle \quad (3.2)$$

$$\langle \text{application} \rangle ::= \langle \text{name} \rangle \langle \text{expression} \rangle^* \quad (3.3)$$

$$\langle \text{function} \rangle ::= \langle \text{name} \rangle \langle \text{pattern} \rangle^* \text{'='} \langle \text{expression} \rangle \quad (3.4)$$

$$\langle \text{pattern} \rangle ::= \langle \text{pattern-value} \rangle (\text{'.'} \langle \text{pattern} \rangle) ? \quad (3.5)$$

$$\langle \text{pattern-value} \rangle ::= \text{'0'} \mid \text{'_'} \mid \text{'('} \langle \text{pattern} \rangle \text{'})' \mid \langle \text{name} \rangle \quad (3.6)$$

$$\langle \text{name} \rangle ::= [\text{'a'}-\text{'z'}] ([\text{'-'}, \text{'a'}-\text{'z'}]^* [\text{'a'}-\text{'z'}]) ? \quad (3.7)$$

¹The extension lends some constructs from regular expressions to achieve a more concise dialect. The extension is described in detail in [Appendix A/21](#).

The term ‘_’ in <pattern-value> is the conventional wildcard operator; it indicates a value that won’t be used by the function declaration, but allows us to keep the same function signature. We define the *signature* of a function as follows:

Definition 3. A function signature in D consists of the function name and the number of parameters it has.

3.2.1 Size

Although the language is already complete, it would prove useful for further analysis to define the notion of *size*, and hence the equality and order of relations on data values. Without further ado, we define the size of a value to be *the number of nodes in the binary tree*.

The equality and order relations are a bit more complicated though, as that there is seemingly no *elegant* way to decide whether one arbitrary binary tree has the same or ..

Hence, the tree 0 has the value 0, the tree 0.0 has the value 1, and the tree 0.0.0 has the value 2 as does its symmetrical equivalent, (0.0).0.

This allows us to define the, otherwise built-in, function `less` in a primitive recursive fashion as follows:

```
less 0 0 = 0
less _ _ 0 = 0
less 0 _ _ = 0.0
less A.B X.Y =

less AR.AL BR.BL = or (and (less AR BR) (less AL BL))
```

This definition indicates that we choose for the empty tree to represent the value *false*, and for the tree 0.0 to represent the value *true*. We’ll keep the definition even more generic, and let the *nonempty* tree represent the value *true*, as shall become useful when we define the higher-order function `if` (§ 3.4.7/14).

Since the values begin at 0 and grow at the rate of 1 ... we can define it as syntactic sugar and use nonnegative integers where ...

In addition to defining the actual data type we need to specify how we’re going to reason about it. Specifically, the questions of equality and order of values constructed in this manner have to be answered.

For all intents and purposes, we can let the *absolute value* of such a tree-structured value be equal to $n - 1$, where n is the number of leafs in the tree. Hence, the tree 0 denotes 0, 0.0 denotes 1, 0.0.0 denotes 2 and so on.

The choice of this data representation yields the following properties for the construction and destruction operators:

Lemma 0.1. Construction of value yields a value strictly greater than either of its constituents. Specifically, the absolute value of the new value is the sum of the absolute values of the constituents.

Lemma 0.2. Destruction of a value yields a pair of values whose absolute values are strictly less than the absolute value of the original value.

3.2.2 Programs

Programs are defined in a conventional functional context and without mutual recursion, namely:

$$\langle \text{program} \rangle ::= \langle \text{function} \rangle^* \langle \text{expression} \rangle \quad (3.8)$$

The order of the function definitions does matter wrt. pattern matching in so far as those defined before are attempted first, if the match fails, the next function with the same signature² is attempted.

²In this case comprising of the name of the function and its arity.

Note, that we let the number of function definitions be zero as an `<expression>` is a valid program as well. More generally, the program can be thought of as a constant function, where the actual `<expression>` simply has access to some predefined functions defined by the function definitions in the program.

3.2.3 Built-in high-order functions

Although D is initially a first-order language, we will ignore that limitation for a bit and define a few higher-order functions to provide some syntactical sugar to the language. Beyond the discussion in this section, these higher-order functions should be regarded as D built-ins.

Branching

In the following definition, the variable names `true` and `false` refer to expressions to be executed in either case.

```
if 0 _ false := false
if _ _ true := true
```

As you can see, we employ the C convention that any value other than 0 is a “truthy” value, and the expression `true` is returned.

Although the call-by-value nature of the language does not allow for short-circuiting the if-statements defined in such a way, this shouldn’t be any impediment to further analysis.

3.2.4 Sample programs

As an illustration of the language syntax, the following program reverses a tree:

```
reverse 0 := 0
reverse left.right := (reverse right).(reverse left)
```

The following program computes the Fibonacci number `n`:

```
fibonacci 0 x y := 0
fibonacci 0.0 x y := y
fibonacci n x y := fibonacci (minus n 0.0) y (add x y)
```

3.3 Semantics

In the following section, the operational semantics of the language D are defined in terms of structured operational semantics[3]. Table 3.2/15 specifies most³ of the syntactical elements used to define the semantic reduction rules below.

In addition to the notation specified in the table, we’ll make use of Haskell-like list comprehension when dealing with lists of elements. For instance, $[e]$ refers to a list of expressions, and $[e'|e]$ refers to a list of expressions that starts with the expression e' and is followed by the list of expressions, e . Also a bit alike Haskell, in both cases above, e is used as both a type and a variable.

3.3.1 Memory

For the sake of an elegant notation, we’ll define the notion of memory as a set of stacks, one for each variable in the program. If a variable n has an empty stack, it is undefined, otherwise the value of the variable (in the present scope) is the value at the top of the corresponding stack.

³The rest is discussed further below and in § 3.5.1/15.

Notation	Description
e	expression
v	value
n	variable name
p	pattern
0	the atom 0
\cdot	'.'
σ	memory
$\sigma[n]$	the value of variable n in memory, returns some v

Table 3.1: Some of the syntactical elements used in the reduction rules for D.

This model of memory allows us to deal with arbitrary scope⁴ in a rather elegant matter, where we simply push a new value onto the corresponding stack when we enter a nested scope and pop off the corresponding stacks when exiting a nested scope. Hence, visiting a nested scope has the following operational semantics wrt. σ :

$$\sigma \longrightarrow \sigma(n) \leftarrow v \longrightarrow \sigma$$

Following the conventions of structured operational semantics, the value of an element, such as σ , does not change throughout a reduction rule. So in the above example, the starting σ is equivalent to the final σ .

3.3.2 Evaluation

D has only one operator, namely '.', which is a right-associative binary operator, hence expressions are evaluated using the following triplet of rules:

$$\frac{\langle e', \sigma \rangle \longrightarrow \langle e'', \sigma \rangle}{\langle e \cdot e', \sigma \rangle \longrightarrow \langle e \cdot e'', \sigma \rangle} \quad (3.9)$$

$$\frac{\langle e, \sigma \rangle \longrightarrow \langle e', \sigma \rangle}{\langle e \cdot v, \sigma \rangle \longrightarrow \langle e' \cdot v, \sigma \rangle} \quad (3.10)$$

$$\langle v \cdot v', \sigma \rangle \longrightarrow \langle v'', \sigma \rangle \quad (\text{where } v'' = v \cdot v') \quad (3.11)$$

Variables

As expressions may contain variable names, we need a way to retrieve the values of variables from memory:

$$\langle n, \sigma \rangle \longrightarrow \langle \sigma[n], \sigma \rangle \quad (3.12)$$

Application

Function application is left-associative and has variable (constant at run time) arity of at least one. The arity of the application depends on the declaration that the function specifier, f , points to. Hence, we begin by evaluating the function specifier itself to some expression λ and a pattern list $[p]$, that is, it's corresponding function declaration:

$$\frac{\langle f, \sigma \rangle \longrightarrow \langle \langle \lambda, [p] \rangle, \sigma \rangle}{\langle \langle f, [e] \rangle, \sigma \rangle \longrightarrow \langle \langle \langle \lambda, [p] \rangle, [e] \rangle, \sigma \rangle} \quad (3.13)$$

Note, in a first-order context λ bares no special meaning, however, the letter is carefully chosen to aid further extension of D to it's higher-order sibling.

⁴Although first-order D can make little use of that.

Followed by evaluation of the pattern and expression lists:

$$\frac{\langle p', \sigma \rangle \longrightarrow \langle p'', \sigma \rangle \wedge \langle e', \sigma \rangle \longrightarrow \langle e'', \sigma \rangle}{\langle p \cdot p', e \cdot e', \sigma \rangle \longrightarrow \langle p \cdot p'', e \cdot e'', \sigma \rangle} \quad (3.14)$$

$$\frac{\langle e', \sigma \rangle \longrightarrow^* \langle 0, \sigma \rangle}{\langle p \cdot 0, e \cdot e', \sigma \rangle \longrightarrow \langle p, e, \sigma \rangle} \quad (3.15)$$

$$\frac{\langle e', \sigma \rangle \longrightarrow^* \langle v', \sigma \rangle}{\langle p \cdot n, e \cdot e', \sigma \rangle \longrightarrow \langle p, e, \sigma(n) \leftarrow v' \rangle} \quad (3.16)$$

We complete application by evaluating λ with the new memory state:

$$\langle \lambda, \sigma([n]) \leftarrow [v] \rangle \longrightarrow \langle v', \sigma \rangle \quad (3.17)$$

Note, we return to the original σ once λ is evaluated.

This is small-step..

$$\frac{\langle n_f, [e], \sigma \rangle \longrightarrow \langle e_f, [p], [e], \sigma \rangle \quad \text{where } \langle e_f, [p] \rangle \text{ is the definition of the function with the name } n_f.}{\langle n_f, [e], \sigma \rangle \longrightarrow \langle v, \sigma \rangle} \quad (3.18)$$

3.4 Data

3.4.1 Representation

The language \mathcal{D} is untyped, and represents all data in terms of *unlabeled ordered binary trees*, henceforth referred to as simply, *binary trees*. Such a tree is recursively defined as follows:

Definition 4. A binary tree is a set that is either empty or contains a single labelless node with a left and right child.

For simplicity, we'll sometimes refer to the empty set as a *leaf*, and the singleton set as a *node*.

To operate on such trees we'll require a few underlying capabilities. Below we'll define these capabilities in general for the purpose of analysis. In § 3.2/5 we'll define the concrete syntax for these in \mathcal{D} .

First and foremost, we need to represent leaves, we do this using the atom *leaf*. Furthermore, we'll define the following operations:

$$\overline{\text{CONSTRUCT}(A_{\text{left}}, A_{\text{right}})} \longrightarrow A \quad \text{If } A_{\text{left}} \text{ and } A_{\text{right}} \text{ are the respective children of } A. \quad (3.19)$$

$$\overline{\text{DESTRUCT}(A)} \longrightarrow \{A_{\text{left}}, A_{\text{right}}\} \quad \text{If } A_{\text{left}} \text{ and } A_{\text{right}} \text{ are the respective children of } A. \quad (3.20)$$

$$\overline{\text{ISNODE}(A)} \longrightarrow \text{true} \quad \text{If } A \text{ is a node.} \quad (3.21)$$

$$\overline{\text{ISNODE}(A)} \longrightarrow \text{false} \quad \text{If } A \text{ is a leaf.} \quad (3.22)$$

$$\frac{\text{ISNODE}(A) \longrightarrow \text{true}}{\text{ISLEAF}(A) \longrightarrow \text{false}} \quad (3.23)$$

$$\frac{\text{ISNODE}(A) \longrightarrow \text{false}}{\text{ISLEAF}(A) \longrightarrow \text{true}} \quad (3.24)$$

Although we won't be needing `ISNODE` and `ISLEAF` once the concrete syntax of \mathcal{D} is in place (since it makes use of pattern matching), it is still worth defining what we'll regard as the values *true* and *false*, for the sake of other functions. We'll adopt a C-like convention:

Definition 5. Any non-leaf binary tree represents the value *true* and any leaf represents the value *false*.

3.4.2 Size

For the purposes of talking about size-change termination, we also need to define the notion of size, and be sure to do so in such a way so that all possible data values are well-founded.

Definition 6. Size of a value in \mathcal{D} is the number of nodes in the tree representing that value.

The “well-foundedness” of D’s data values, given such a definition can be argued for by equating the definition to a mapping of D’s data values onto the natural numbers. This would imply that we can define the relation $<$ on D’s data values, which we know to be well-founded.

This is a bijection

We start by formally proving that Definition 6/10 yields a many-to-one mapping of D’s data values to the natural numbers.

First, we prove, by induction, that any natural number can be represented in D:

Proof.

Base A *leaf* has no nodes, and hence represents the value 0.

Assumption If we can represent the $n \in \mathbb{N}$ in D, then we can also represent the number $n + 1 \in$.

Induction Let n be represented by some binary tree A , then $n + 1$ can be represented by $\text{CONSTRUCT}(\text{leaf}, A)$.

□

Second, we prove, also by induction, that any value D has a representation in \mathbb{N} .

Proof.

Base A *leaf* has no nodes, and hence corresponds to the value 0.

Assumption 1. If the binary tree A has a representation $n \in \mathbb{N}$, then $|\text{CONSTRUCT}(\text{leaf}, A)| \equiv n + 1$ and $|\text{CONSTRUCT}(A, \text{leaf})| \equiv n + 1$.
2. If the binary tree A has a representation $n \in \mathbb{N}$, and the binary tree B has a representation $m \in \mathbb{N}$, then $|\text{CONSTRUCT}(A, B)| \equiv n + m + 1$ and $|\text{CONSTRUCT}(B, A)| \equiv n + m + 1$.

Induction By definition of the binary operation CONSTRUCT ,

$$|A| = 1 + |A_{\text{left}}| + |A_{\text{right}}|$$

Hence, the assumptions must hold.

□

Definition 6/10 *almost* allows us to devise an algorithm to compare the sizes of data values. The problem withstanding is that two different values can have rather diverging tree representations. Hence, comparing them, using only the operations defined in § 3.4.1/10, is seemingly impossible unless we initially, or along the way, transform the binary trees being compared into some sort of a *standard representation*. We’ll define this representation, recursively, as follows:

Definition 7. A binary tree in *standard representation* is a binary tree that either is a *leaf* or a node having a *leaf* as it’s left child and a binary tree in *standard representation* as it’s right child.

Intuitively, a binary tree in standard representation is just a tree that only descends along the right side. Comparing the sizes of two trees in this representation is just a matter of walking the descending in the two trees simultaneously, until one of them, or both, bottom out. If there is a tree that bottoms out strictly before another, that is the lesser tree by Definition 6/10.

3.4.3 LESS(A, B)

Assuming that we've already defined a procedure NORMALIZE for transforming an arbitrary tree into its standard representation (which we'll do further below), we can define the procedure LESS(A, B) for determining whether the value of the binary tree A is strictly less than the value of binary tree B , as follows:

```
LESS( $A, B$ )
1  NORMALIZEDLESS(NORMALIZE( $A$ ), NORMALIZE( $B$ ))

NORMALIZEDLESS( $A, B$ )
1  if ISLEAF( $A$ )
2      return ISNODE( $A$ )
3  if ISLEAF( $B$ )
4      return false

5  { $\_A_{right}$ } = DESTRUCT( $A$ )
6  { $\_B_{right}$ } = DESTRUCT( $B$ )

7  return NORMALIZEDLESS( $A_{right}, B_{right}$ )
```

Correctness

Proof. Given Definition 7/11, and the assumption that NORMALIZE(A) computes the standard representation of A , we know the following:

1. $|A| \equiv |\text{NORMALIZE}(A)|$.
2. We'll walk through all the nodes if we perform a recursive right-child-walk starting at A .
3. The same holds for B .

It is also easy to see from lines 1:4 that NORMALIZEDLESS stops as soon as we reach the "bottom" of either A or B .

Given Definition 6/10, $A < B$ iff it bottoms out before B , that is, we reach an instance of the recursion where both $\text{IsLeaf}(A)$ and $\text{IsNode}(B)$ hold. In all other cases $A \geq B$, the cases specifically are:

- $\text{IsLeaf}(A)$ and $\text{IsLeaf}(B)$, then $|A| \equiv |B|$.
- $\text{IsNode}(A)$ and $\text{IsLeaf}(B)$, then $|A| > |B|$

Last but not least, due to all data values being finite, eventually one of the trees does bottom out. □

Time complexity

Given that the binary trees A and B are in standard representation when we enter the auxiliary procedure, NORMALIZEDLESS, it is fairly easy to get an upper bound on the running time of NORMALIZEDLESS itself.

Indeed, the running time of NORMALIZEDLESS itself is $O(\text{MAX}(|A|, |B|))$, since we just walk down the trees until one of them bottoms out.

We haven't yet defined the procedure NORMALIZE yet. Hence, the only thing that we can say about the running time of LESS in general is that it is $O(\text{NORMALIZE}(A) + \text{NORMALIZE}(B) + \text{MAX}(|A|, |B|))$.

Space complexity

Coming soon..

3.4.4 NORMALIZE(A)

To complete the definition and analysis of LESS we need to define and analyze NORMALIZE. We do this below.

NORMALIZE(A)

1 NORMALIZEAUXILIARY($A, leaf, leaf$)

NORMALIZEAUXILIARY($A, A', A_{normalized}$)

```

1  if ISNODE( $A$ )
2       $A_{normalized} = \text{CONSTRUCT}(leaf, A_{normalized})$ 
3  else
4      return  $leaf$ 

5   $\{A_{left}, A_{right}\} = \text{DESTRUCT}(A)$ 
6  if ISNODE( $A_{left}$ )
7       $A' = \text{CONSTRUCT}(A_{left}, A')$ 
8  if ISNODE( $A_{right}$ )
9       $A = A_{right}$ 
10 else
11     if ISNODE( $A'$ )
12          $\{A, A'\} = \text{DESTRUCT}(A')$ 
13     else
14         return  $A_{normalized}$ 
15 return NORMALIZE( $A, A', A_{normalized}$ )

```

Correctness

Coming soon..

Time complexity

Coming soon..

Space complexity

Coming soon..

3.4.5 Size

Although the language is already complete, it would prove useful for further analysis to define the notion of *size*, and hence the equality and order of relations on data values. Without further a-do, we define the size of a value to be *the number of nodes in the binary tree*.

The equality and order relations are a bit more complicated though, as that there is seemingly no *elegant* way to decide whether one arbitrary binary tree has the same or ..

Hence, the tree 0 has the value 0, the tree 0.0 has the value 1, and the tree 0.0.0 has the value 2 as does it's symmetrical equivalent, (0.0).0.

This allows us to define the, otherwise built-in, function less in a primitive recursive fashion as follows:

```

less 0 0 = 0
less _ _ 0 = 0
less 0 _ _ = 0.0
less A.B X.Y =

```

```

less AR.AL BR.BL = or (and (less AR BR) (less AL BL))

```

This definition indicates that we choose for the empty tree to represent the value *false*, and for the tree `0.0` to represent the value *true*. We'll keep the definition even more generic, and let the *nonempty* tree represent the value *true*, as shall become useful when we define the higher-order function `if` (§ 3.4.7/14).

Since the values begin at 0 and grow at the rate of 1 ... we can define it as syntactic sugar and use nonnegative integers where ...

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The choice of this data representation yields the following properties for the construction and destruction operators:

Lemma 0.3. *Construction of value yields a value strictly greater than either of it's constituents. Specifically, the absolute value of the new value is the sum of the absolute values of the constituents.*

Lemma 0.4. *Destruction of a value yields a pair of values who's absolute values are strictly less than the absolute value of the original value.*

3.4.6 Programs

Programs are defined in a conventional functional context and without mutual recursion, namely:

$$\langle \text{program} \rangle ::= \langle \text{function} \rangle^* \langle \text{expression} \rangle \quad (3.25)$$

The order of the function definitions does matter wrt. pattern matching in so far as those defined before are attempted first, if the match fails, the next function with the same signature⁵ is attempted.

Note, that we let the number of function definitions be zero as an `<expression>` is a valid program as well. More generally, the program can be thought of as a constant function, where the actual `<expression>` simply has access to some predefined functions defined by the function definitions in the program.

3.4.7 Built-in high-order functions

Although D is initially a first-order language, we will ignore that limitation for a bit and define a few higher-order functions to provide some syntactical sugar to the language. Beyond the discussion in this section, these higher-order functions should be regarded as D built-ins.

Branching

In the following definition, the variable names `true` and `false` refer to expressions to be executed in either case.

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As you can see, we employ the C convention that any value other than 0 is a "truthy" value, and the expression `true` is returned.

Although the call-by-value nature of the language does not allow for short-circuiting the if-statements defined in such a way, this shouldn't be any impediment to further analysis.

⁵In this case comprising of the name of the function and it's arity.

3.4.8 Sample programs

As an illustration of the language syntax, the following program reverses a tree:

```
reverse 0 := 0
reverse left.right := (reverse right).(reverse left)
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The following program computes the Fibonacci number n :

```
fibonacci 0 x y := 0
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fibonacci n x y := fibonacci (minus n 0.0) y (add x y)
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3.5 Semantics

In the following section, the operational semantics of the language D are defined in terms of structured operational semantics[3]. Table 3.2/15 specifies most⁶ of the syntactical elements used to define the semantic reduction rules below.

In addition to the notation specified in the table, we'll make use of Haskell-like list comprehension when dealing with lists of elements. For instance, $[e]$ refers to a list of expressions, and $[e'|e]$ refers to a list of expressions that starts with the expression e' and is followed by the list of expressions, e . Also a bit alike Haskell, in both cases above, e is used as both a type and a variable.

Notation	Description
e	expression
v	value
n	variable name
p	pattern
0	the atom 0
$.$	'.'
σ	memory
$\sigma[n]$	the value of variable n in memory, returns some v

Table 3.2: Some of the syntactical elements used in the reduction rules for D .

3.5.1 Memory

For the sake of an elegant notation, we'll define the notion of memory as a set of stacks, one for each variable in the program. If a variable n has an empty stack, it is undefined, otherwise the value of the variable (in the present scope) is the value at the top of the corresponding stack.

This model of memory allows us to deal with arbitrary scope⁷ in a rather elegant matter, where we simply push a new value onto the corresponding stack when we enter a nested scope and pop off the corresponding stacks when exiting a nested scope. Hence, visiting a nested scope has the following operational semantics wrt. σ :

$$\sigma \longrightarrow \sigma(n) \leftarrow v \longrightarrow \sigma$$

Following the conventions of structured operational semantics, the value of an element, such as σ , does not change throughout a reduction rule. So in the above example, the starting σ is equivalent to the final σ .

⁶The rest is discussed further below and in § 3.5.1/15.

⁷Although first-order D can make little use of that.

3.5.2 Evaluation

\mathcal{D} has only one operator, namely ‘ \cdot ’, which is a right-associative binary operator, hence expressions are evaluated using the following triplet of rules:

$$\frac{\langle e', \sigma \rangle \longrightarrow \langle e'', \sigma \rangle}{\langle e \cdot e', \sigma \rangle \longrightarrow \langle e \cdot e'', \sigma \rangle} \quad (3.26)$$

$$\frac{\langle e, \sigma \rangle \longrightarrow \langle e', \sigma \rangle}{\langle e \cdot v, \sigma \rangle \longrightarrow \langle e' \cdot v, \sigma \rangle} \quad (3.27)$$

$$\langle v \cdot v', \sigma \rangle \longrightarrow \langle v'', \sigma \rangle \quad (\text{where } v'' = v \cdot v') \quad (3.28)$$

Variables

As expressions may contain variable names, we need a way to retrieve the values of variables from memory:

$$\langle n, \sigma \rangle \longrightarrow \langle \sigma[n], \sigma \rangle \quad (3.29)$$

Application

Function application is left-associative and has variable (constant at run time) arity of at least one. The arity of the application depends on the declaration that the function specifier, f , points to. Hence, we begin by evaluating the function specifier itself to some expression λ and a pattern list $[p]$, that is, it's corresponding function declaration:

$$\frac{\langle f, \sigma \rangle \longrightarrow \langle \langle \lambda, [p] \rangle, \sigma \rangle}{\langle \langle f, [e] \rangle, \sigma \rangle \longrightarrow \langle \langle \langle \lambda, [p] \rangle, [e] \rangle, \sigma \rangle} \quad (3.30)$$

Note, in a first-order context λ bares no special meaning, however, the letter is carefully chosen to aid further extension of \mathcal{D} to it's higher-order sibling.

Followed by evaluation of the pattern and expression lists:

$$\frac{\langle p', \sigma \rangle \longrightarrow \langle p'', \sigma \rangle \wedge \langle e', \sigma \rangle \longrightarrow \langle e'', \sigma \rangle}{\langle p \cdot p', e \cdot e', \sigma \rangle \longrightarrow \langle p \cdot p'', e \cdot e'', \sigma \rangle} \quad (3.31)$$

$$\frac{\langle e', \sigma \rangle \longrightarrow^* \langle 0, \sigma \rangle}{\langle p \cdot 0, e \cdot e', \sigma \rangle \longrightarrow \langle p, e, \sigma \rangle} \quad (3.32)$$

$$\frac{\langle e', \sigma \rangle \longrightarrow^* \langle v', \sigma \rangle}{\langle p \cdot n, e \cdot e', \sigma \rangle \longrightarrow \langle p, e, \sigma(n) \leftarrow v' \rangle} \quad (3.33)$$

We complete application by evaluating λ with the new memory state:

$$\langle \lambda, \sigma([n]) \leftarrow [v] \rangle \longrightarrow \langle v', \sigma \rangle \quad (3.34)$$

Note, we return to the original σ once λ is evaluated.

This is small-step..

$$\frac{\langle n_f, [e], \sigma \rangle \longrightarrow \langle e_f, [p], [e], \sigma \rangle \quad \text{where } \langle e_f, [p] \rangle \text{ is the definition of the function with the name } n_f.}{\langle n_f, [e], \sigma \rangle \longrightarrow \langle v, \sigma \rangle} \quad (3.35)$$

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Bibliography

- [1] P. Naur (ed.), Revised Report on the Algorithmic Language ALGOL 60; CACM, Vol. 6, p. 1; The Computer Journal, Vol. 9, p. 349; Num. Math., Vol. 4, p. 420. (1963); Section 1.1.
- [2] A. M. Turing, On computable numbers with an application to the Entscheidungsproblem; Proceedings of the London Mathematical Society, 42(2):230-265, (1936).
- [3] Gordon D. Plotkin, A Structural Approach to Operational Semantics, Journal of Logic and Algebraic Programming (2004) Volume: 60-61, Issue: January, Publisher: Citeseer, Pages: 17-139.

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Appendix A

Extended-BNF

This report makes use of an extended version of the Backus-Naur form (BNF). This appendix is provided to cover the extensions employed in the report. This is done because there is seemingly no universally acknowledged extension, unlike there is a universally acknowledged Backus-Naur form, namely the one used in the ALGOL 60 Reference Manual[1].

A.1 What's in common with the original BNF

The following parts are in-common with the original Backus-Naur form:

Construct	Description
< ... >	A metalinguistic variable, aka. a nonterminal.
::=	Definition symbol
	Alternation symbol

Table A.1: Constructs in common with the original BNF.

In the original BNF, everything else represents itself, aka. a terminal. This is not preserved in this extension – all terminals are encapsulated into single quotes.

A.2 Constructs borrowed from regular expressions.

The use of single quotes around all terminals allows us to give characters such as (,),], *, +, and * special meaning, namely:

Construct	Meaning
(...)	Entity group
[...]	Character group
-	Character range
*	0-∞ repetition
+	1-∞ repetition
?	0-1 repetition

Table A.2: Constructs borrowed from regular expressions.

An entity group is a shorthand for an auxiliary nonterminal declaration. This means, for instance, that using the alternation symbol within it would mean an alternation of entity sequences within the entity group rather than the entire declaration that contains the entity group.

A character group may only contain single character terminals and an alternation of the terminals is implied from their mere sequence. It is identical to an auxiliary single character nonterminal declaration. A character range binary operator can be used to shorten a given character group, e.g. [`'a'` - `'z'`] implies the list of characters from `'a'` to `'z'` in the ASCII table. Moreover, a character range is the only operator allowed in a character group.

Applying the repetition operators to either the closing brace of an entity group or the closing bracket of a character group has the same effect as applying the repetition operator to their respective hypothetical auxiliary declarations.

A.3 Nonterminals as sets and conditional declarations

Another extension to the original BNF is the ability to use nonterminals as sets in declaration conditions. For example, if the two nonterminals, `<type-name>` and `<constructor-name>`, are both declared in terms of the `<literal>` nonterminal, but type names and constructor names should not intersect in a given program, then we can append the following condition to one or both declarations:

$$\text{s.t. } \langle \text{type-name} \rangle \cap \langle \text{constructor-name} \rangle \equiv \emptyset$$

Where the shorthand s.t. stands for “such that”. This implies that the nonterminals `<type-name>` and `<constructor-name>` represent the sets of character sequences that end up associated with the respective nonterminals for any given program, and can be used in conjunction with regular set notation.