

# Aspects of efficiency in functional programming languages

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## Part I

# Compilers and languages

# Chapter 1

## Programming languages

Computers are devices which read a well-defined, finite sequence of simple instructions and emit a result. In theoretical analysis of computers, models have been developed to understand and prove properties. A finite sequence of instructions fed to a computer is called an *algorithm*, which is the language of high level computation [Cop97]. In modern encodings of algorithms or programs, “high level” languages are used instead of the computational models. Such languages are then translated into instructions that often are much closer to a computational model. The process of translating programs into computer instructions is called *compiling*, or *transpiling* if the program is first translated into another “high level” language.

For the purpose of this dissertation, a simple programming language has been implemented to illustrate the concepts in detail. The language transpiles to *untyped lambda calculus*. For the remainder, the language will be referred to as  $L$ .

## 1.1 Untyped lambda calculus

The *untyped lambda calculus* is a model of computation developed by Alonzo Church [Chu36]. The untyped lambda calculus is a simple tangible language of just three terms.

$$x \tag{1.1}$$

$$\lambda x.E \tag{1.2}$$

$$YE \tag{1.3}$$

Equation 1.2 displays a lambda *abstraction* which essentially is a function that states “given some  $x$  compute  $E$ ” where  $E$  is another one of the three terms in which  $x$  may occur. The *variable* (Equation 1.1) is a reference to some value introduced by an abstraction. A variable is a reference to another lambda abstraction. In the untyped lambda calculus there is also the notion of *context* which simply means where in a lambda expression something is computed. Context is important when discussing *free* and *bound* variables as whether a variable is free or bound is decided by the context. Free variables are determined by Equation 1.4, Equation 1.5 and Equation 1.6.

$$Free(x) = \{x\} \tag{1.4}$$

$$Free(\lambda x.E) = Free(E) \setminus \{x\} \tag{1.5}$$

$$Free(YE) = Free(Y) \cup Free(E) \tag{1.6}$$

**Example 1.1.1.**

$$\lambda x.\lambda y.x \tag{1.7}$$

In Equation 1.7  $x$  can appear both free and bound based on the context. If the context is  $\lambda y.x$  then  $x$  appears free but given the whole expression  $x$  appears bound.

In Equation 1.3 the *application* term is displayed. An application of two terms can be interpreted as substituting the variable in the left abstraction  $Y$  with the right term  $E$ . It is also common to introduce the *let binding* to the untyped lambda calculus which will be further discussed when introducing typing in section 2.1.

**Example 1.1.2.** Let  $Y$  be  $\lambda x.T$  and  $E$  be  $z$  then  $YE$  is  $(\lambda x.T)z$ . Furthermore substituting  $x$  for  $E$  such that  $Y$  becomes  $T[x := E]$ . Since  $E = z$  then substitute  $E$  for  $z$  such that  $T[x := z]$  read as “Every instance of  $x$  in  $T$  should be substituted by  $z$ ”.

**Remark 1.1.1.** Substituting lambda terms is a popular method of evaluating lambda calculus programs. Languages like Miranda Clean and general purpose evaluation programs like the G-machine implement *combinator graph rewriting* which is similar and will be introduced in subsection 3.2.1.

A remarkable fact about the untyped lambda calculus is that it is turing complete; any algorithm that can be evaluated by a computer can be encoded in the untyped lambda calculus. The turing completeness of the untyped lambda calculus can be realized by modelling numerics, boolean logic and recursion with the *Y-combinator*. Church encoding is the encoding of numerics, arithmetic expressions and boolean logic [Chu85]. Church encoding may prove the power of the untyped lambda calculus but has terrible running time for numerics since to represent some  $n \in \mathbb{Z}$  it requires  $n$  applications. For the remainder of the dissertation ordinary arithmetic expressions are written in traditional mathematics. The simplicity of lambda calculus makes it an excellent language to transpile to which is a common technique.

## 1.2 Translation to lambda calculus

High level languages associated with lambda calculus are often also very close to it. The *L* language is very close to the untyped lambda calculus. See two equivalent programs Listing 1.1 and Listing 1.2 that both add an *a* and a *b*.

Listing 1.1: Add function in lambda calculus

```
1 (λadd.E)(λa.λb.a + b)
```

Listing 1.2: Add function in L

```
1 fun add a b = a + b;
```

Notice that in Listing 1.1 the term *E* is left undefined, *E* is “the rest of the program in this scope”. If the program were to apply 1 and 2 to add the resulting program in L would be Listing 1.4 and in the untyped lambda calculus it would be Listing 1.3.

Listing 1.3: Add function in lambda calculus

```
1 (λadd.add 1 2)(λa.λb.a + b)
```

Listing 1.4: Add function applied

```
1 fun add a b = a + b;
2 add 1 2;
```

### 1.2.1 Scoping

Notice that Listing 1.1 must bind the function name “outside the rest of the program” or more formally in an outer scope. In a traditional program

such as Listing 1.5 functions must be explicitly named to translate as in the above example.

Listing 1.5: A traditional program

```
1 fun add a b = a + b;
2 fun sub a b = a - b;
3 sub (add 10 20) 5;
```

Listing 1.6: An order dependent program

```
1 fun sub a b = add a (0 - b);
2 fun add a b = a + b;
3 sub (add 10 20) 5;
```

Notice that there are several problems such as the order of which functions are defined may alter whether the program is correct or not. For instance the program defined in Listing 1.6 would not translate into a valid program, it would translate into Listing 1.7. The definition of `sub` is missing a reference to the `add` function.

Listing 1.7: Add function in lambda calculus

```
1 (λsub.λadd. sub (add 10 20) 5)
2   (λa.λb.a + b)
3   (λa.λb.add a (0 - b))
```

*lambda lifting* is a technique where free variables (section 1.1) are explicitly parameterized [Joh85]. This is exactly the problem in Listing 1.7 which has the lambda lifted solution Listing 1.8.

Listing 1.8: Order dependent

```
1 (λsub.λadd. sub add (add 10 20) 5)
2   (λa.λb.a + b)
3   (λadd.λa.λb.add a (0 - b))
```

As it will turn out this will also enables complicated behaviour such as *mutual recursion*.

Moreover lambda lifting also conforms to “traditional” scoping rules. *Variable shadowing* occurs when there exists  $1 < \text{reachable variables}$  of the same name but the “nearest” in regard to scope distance is chosen. Effectively other variables than the one chosen are *shadowed*. Variable shadowing is an implied side-effect of using using lambda calculus. The function `f` in Listing 1.9 yields 12.

Listing 1.9: Scoping rules in programming languages

```
1 let x = 22;
2 let a = 10;
```



```

3 fun f =
4   let x = 2;
5   a + x;

```

### 1.2.2 Recursion

Reductions in mathematics and computer science are one of the principal methods used for developing beautiful equations and algorithms.

Listing 1.10: Infinite program

```

1 fun f n =
2   if (n == 0) n
3   else if (n == 1) n + (n - 1)
4   else if (n == 2) n + ((n - 1) + (n - 2))
5   ...

```

Listing 1.10 defines a function **f** that in fact is infinite. In the untyped lambda calculus there are not any of the three term types that define infinite functions or abstractions, at first glance. Instead of writing an infinite function the question is rather how can a reduction be performed on this function such that it can evaluate *any* case of **n**?

Listing 1.11: Recursive program

```

1 fun f n =
2   if (n == 0) n
3   else n + (f (n - 1))

```

Listing 1.11 defines a recursive variant of **f** it is a product of the reduction in Equation 1.8.

$$n + (n - 1) \cdots + 0 = \sum_{k=0}^n k \quad (1.8)$$

Since the untyped lambda calculus is turing complete or rather if one were to show it were it must also realize algorithms that are recursive or include loops (the two of which are equivalent in expressiveness).

Listing 1.12: Recursive function

```

1 (λf.E) (λn.if (n == 0) (n) (n + (f (n - 1))))

```

The naive implementation of a recursive variant will yield an unsolvable problem which in fact is an infinite problem. In Listing 1.12 when **f** is applied recursively it must be referenced while it is “being constructed”. Substituting *f* with its implementation in Listing 1.13 will yield the same problem again but at one level deeper. The **if** function takes a condition,

the body in case of the condition being true and the body in case of the condition being false.

Listing 1.13: Recursive function `f` substituted

```

1  (λf.E)
2  (λn.if (n == 0) (n) (n + (
3      (λn.if (n == 0) (n) (n + (f (n - 1))))
4      (n - 1)
5  )))

```

One could say that the problem is now recursive. Recall that lambda lifting (subsection 1.2.1) is the technique of explicitly parameterizing outside references. Convince yourself that `f` lives in the scope above its own body such that when referencing `f` from within `f`, `f` should be parameterized as in Listing 1.14 such that it translates to Listing 1.15.

Listing 1.14: Explicitly passing recursive function

```

1  fun f f n =
2      if (n == 0) n
3      else n + (f f (n - 1))

```

Listing 1.15: Explicitly passing recursive function in the lambda calculus

```

1  (λf.E)(λf.λn. if (n == 0) (n) (n + (f f (n - 1))))

```

The initial invocation of `f` must involve `f` such that it becomes `f f n`. The *Y-combinator* an implementation of a fixed-point combinator in Equation 1.9 is the key to realize that the untyped lambda calculus can implement recursion. Languages with functions and support binding functions to parameters can implement recursion with the Y-combinator.

$$\lambda f.(\lambda x.f(xx))(\lambda x.f(xx)) \quad (1.9)$$

Implementing mutual recursion is an interesting case of lambda lifting and recursion in untyped lambda calculus.

Listing 1.16: Mutual recursion

```

1  fun g x = f x;
2  fun f x = g x;

```

Notice in Listing 1.16 that `g` requires `f` to be lifted and `f` requires `g` to be lifted. If a translation “pessimistically” lifts all definitions from the above scope then all required references exist in lexical scope.

Languages have different methods of introducing recursion some of which

have very different implications especially when considering types. For instance OCaml has the `let rec` binding to introduce recursive definitions. The `rec` keyword indicates to the compiler that the binding should be able to “see itself” (??).

## 1.3 High level abstractions

The lambda calculus is a powerful language that can express any algorithm. Expressiveness does not necessarily imply ergonomics or elegance, in fact encoding moderately complicated algorithms in lambda calculus becomes quite messy. Many high level techniques exist to model abstractions in tangible concepts.

### 1.3.1 Algebraic data types

Algebraic data types are essentially a combination of disjoint unions, tuples and records. Algebraic data types are closely related to types thus require some type theory to fully grasp. Types are explored more in depth in ??.

Listing 1.17: List algebraic data type

```
1 type List a =  
2   | Nil  
3   | Cons a (List a)  
4 ;
```

Listing 1.17 is an implementation of a linked list. The list value can either take the type of `Nil` indicating an empty list, or it can take the type of `Cons` indicating a pair of type *a* and another list. The list implementation has two constructors and one type parameter. The type parameter *a* of the list algebraic data type defines a *polymorphic type*; *a* can agree on any type, it is universally quantified  $\forall a$ . `Cons a (List a)`. The two constructors `Nil` and `Cons` both create a value of type `List a` once instantiated.

Listing 1.18: List instance and match

```
1 let l = Cons 1 (Cons 2 (Cons 3 Nil));  
2  
3 match l  
4   | Nil -> 0;  
5   | Cons n _ -> n;  
6 ;
```

Once a value is embedded into an algebraic data type such as a list it must be extractable to be of any use. Values of algebraic data types are extracted and analysed with *pattern matching*. Pattern match comes in many forms,

notably it allows one to define a computation based on the type an algebraic data type instance realizes (Listing 1.18).

### Scott encoding

Pattern matching strays far from the simple untyped lambda calculus, but can in fact be encoded into it. The *scott encoding* (Equation 1.10) is a technique that describes a general purpose framework to encode algebraic data types into lambda calculus [Sco62]. Considering an algebraic data type instance as a function which accepts a set of “handlers” allows the encoding into lambda calculus. The scott encoding specifies that constructors should now be functions that are each parameterized by the constructor parameters  $x_1 \dots x_{A_i}$  where  $A_i$  is the arity of the constructor  $i$ . Additionally each of the constructor functions return a  $n$  arity function, where  $n$  is the cardinality of the set of constructors. Of the  $n$  functions, the constructor parameters  $x_1 \dots x_{A_i}$  are applied to the  $i$ ’th “handler”  $c_i$ . These encoding rules ensure that the “handler” functions are provided uniformly to all instances of the algebraic data type.

$$\lambda x_1 \dots x_{A_i}. \lambda c_1 \dots c_n. c_i x_1 \dots x_{A_i} \quad (1.10)$$

**Example 1.3.1.** The `List` algebraic data type in Listing 1.17 has two constructors, `Nil` with the constructor type Equation 1.11 and `Cons` with the constructor type Equation 1.12. Equation 1.11 is in fact also the type of `List` once instantiated, effectively treating partially applied functions as data.

$$b \rightarrow (a \rightarrow \text{List } a \rightarrow b) \rightarrow b \quad (1.11)$$

$$(a \rightarrow \text{List } a \rightarrow b) \rightarrow b \rightarrow (a \rightarrow \text{List } a \rightarrow b) \rightarrow b \quad (1.12)$$

Listing 1.19: List algebraic data type implementation

```

1 fun cons x xs =
2   fun c _ onCons = onCons x xs;
3   c;
4
5 fun nil =
6   fun c onNil _ = onNil;
7   c;
```

Encoding the constructors in  $L$  yields the functions defined in Listing 1.19. Pattern matching is but a matter of applying the appropriate handlers. In Listing 1.20.

Listing 1.20: Example of scott encoded list algebraic data type

```

1 | let l = cons 1 (cons 2 (cons 3 nil));
2 |
3 | fun head x _ =
4 |   x;
5 |
6 | l head 0;

```

Efficiency can be a bit tricky in lambda calculus as it is at the mercy of implementation. A common method of considering efficiency is counting  $\beta$ -reduction since they evaluate to function invocations. The  $\beta$ -reduction is a substitution which substitutes an application where the left side is an abstraction in which the bound variable is substituted with the right side term (Equation 1.13).

$$\beta_{red}((\lambda x.T)E) = T[x := E] \quad (1.13)$$

It should be clear that invoking a  $n$  arity function will take  $n$  applications. In the case of a scott encoded algebraic data types the largest term in regard to complexity is either the size of the set of “handler” functions or the “handler” function with most parameters. The time to evaluate pattern match is thus  $O(\max_i(c_i + A_i))$ .

## Chapter 2

# Typing and validation

Automatic validation is one of many reasons to use computers for solving various tasks including writing new computer programs. Spellchecking is a common and trivial instance of an input validation algorithm.

### 2.1 Types and validation

The spell checking equivalent for computer programs could be type checking; a subproblem of validating a programmer’s intuition of a program’s intent. Types also have other properties than simply validating they can in fact be related to theorems to which an implementation is the proof [How80].

Listing 2.1: Head implementation

```
1 fun head l: List a → a =  
2   match l  
3     | Cons x _ → x;  
4     | Nil → ?;  
5   ;
```

For instance, consider the implementation of the function with type `List a → a` in Listing 2.1. A total implementation of the function cannot exist.

The type system for the *L* language will be the Hindley-Milner type system [Hin69; Mil78].

#### 2.1.1 The language of types

Before delving into types, the lambda calculus defined in section 1.1 must be augmented with the *let expression* (Equation 2.1).

$$\text{let } x = Y \text{ in } E \tag{2.1}$$

It should be noted that the let binding can be expressed by abstraction and application (Equation 2.2).

$$(\lambda x.E)(Y) \quad (2.2)$$

The let expression has a nice property that will become apparent later when typing rules are introduced.

Types are an artificial layer atop of a program just as spell checking is an artificial layer atop text. There are two variants of types in the Hindley-Milner type system, the *monotype* and the *polytype*. A monotype is either a type variable, an abstraction of two monotypes or an application of a type constructor (Equation 2.3).

$$mono\ \tau = a \mid \tau \rightarrow \tau \mid C\tau_1 \dots \tau_n \quad (2.3)$$

*Atoms* are terminal terms in a formula and are expressed either by type variable  $a$  or  $C$  with no type parameters. The application term of the monotype is dependent on the primitive types of the programming language. The types  $\tau_1 \dots \tau_n$  are monotype parameters required to construct some type  $C$ . In  $L$  the set of type constructors are  $\{\mathbf{Int}, \mathbf{Bool}\} \cup \mathbf{ADT}$ .  $\mathbf{Int}$  and  $\mathbf{Bool}$  are type constructors of arity 0 thus only have one instantiation and are atomic. The set of constructors  $\mathbf{ADT}$  encapsulates the set of program defined algebraic data structures (??).

**Example 2.1.1.** Let  $\mathbf{ADT} = \{\mathbf{List}\}$  where  $\mathbf{List}$  is defined as in Listing 1.17. The *type constructor* (not to be confused for constructors like  $\mathbf{Cons}$  or  $\mathbf{Nil}$ ) for  $\mathbf{List}$  has the signature  $\mathbf{a} \rightarrow \mathbf{List}\ \mathbf{a}$  stating that if supplied with some type  $\mathbf{a}$  it constructs a type of  $\mathbf{List}\ \mathbf{a}$  (effectively containing the provided type). The type  $\mathbf{List}$  is a type constructor with one type parameter  $\mathbf{a}$ .

$\perp$  denotes falsity, in type systems a value of this type can never exist since that in itself would disprove the program. It is common in programming languages with strong type systems to let thrown exceptions be of type  $\perp$  since it adheres to every type and indicates that the program is no longer running, since no instance of  $\perp$  can exist.  $\top$  denotes truth, in type systems every type is a supertype of  $\top$ .  $\top$  is in practice only used to model side effects, since not all side effects return useful values. In programming languages with side effects  $\perp$  and  $\top$  are considerably more useful than in pure programming languages.

A polytype is a polymorphic type (Equation 2.4).

$$poly\ \sigma = \tau \mid \forall a.\sigma \quad (2.4)$$

Polymorphic types either take the shape of a type variable or introduce a type which all types  $a$  adhere to. This does not necessarily include *all* types since the **Gen** rule of Figure 2.2 constrains the domain that  $a$  ranges

over to contain only type variables that are not free. Many types may adhere to a polymorphic type but polymorphic types do not adhere to any type other than polymorphic types. The concept of adherence in types is commonly called *subtyping*. Every subtype is a *at least* an implementation of it's supertype. Since this concept can be difficult to grasp from just text, observe Figure 2.1. Note that  $\sigma$  is controversial to introduce to the type



Figure 2.1: The type hierarchy of Hindley-Milner.

hierarchy and has only been so to illustrate the point of subtyping.  $\sigma$  is but a mechanism to prove type systems,  $\sigma$  is never a specific type.

**Remark 2.1.1.** An important implementation detail which should be noted is that of the polymorphic type. Polymorphic types can be regarded as being a pair of bound types and monotype. Instead of keeping track of what types cannot occur, keeping track of the ones than can occur simplifies the implementation. This representation is convenient for the **Gen** rule.

A principal component of typing in Hindley-Milner is the *environment*. The environment  $\Gamma$  is a set of pairs of variable names and polytype (Equation 2.5).  $\Gamma \vdash x : \sigma$  implies a *typing judgment*, meaning that given  $\Gamma$ , the variable  $x$  can take the type  $\sigma$ .

**Remark 2.1.2.** Judging a type does not necessarily mean that the judged type is the only type that  $x$  may take, it states that it is one *possible* type that  $x$  may take. The property of taking multiple possible types is what allows polymorphism. This is made more apparent in Example 2.2.2 where `id` may take the type of either  $\forall a. a \rightarrow a$ ,  $\text{Int} \rightarrow \text{Int}$  or  $\forall a. (a \rightarrow a) \rightarrow (a \rightarrow a)$ .

$$\Gamma = \epsilon \mid \Gamma, x : \sigma \quad (2.5)$$

Like in the untyped lambda calculus, types also have notions of free and bound type variables. Bound type variables are ones that explicitly



have been introduced to the type system by either let or abstraction in the context of some expression. Type variables are bound when they have been introduced by a quantification or exist in the environment.

$$\text{free}(a) = \{a\} \quad (2.6)$$

$$\text{free}(C\tau_1 \dots \tau_n) = \bigcup_{i=1}^n \text{free}(\tau_i) \quad (2.7)$$

$$\text{free}(\tau_1 \rightarrow \tau_2) = \text{free}(\tau_1) \cup \text{free}(\tau_2) \quad (2.8)$$

$$\text{free}(\Gamma) = \bigcup_{x:\sigma \in \Gamma} \text{free}(\sigma) \quad (2.9)$$

$$\text{free}(\forall a.\sigma) = \text{free}(\sigma) - \{a\} \quad (2.10)$$

**Example 2.1.2.** Consider the type for the function `fst` in Listing 2.3.

Listing 2.2: First function

```
1 fun fst a b: ∀A.∀B.A → B → A = a;
```

Listing 2.3: First function in lambda calculus

```
1 let fst = λa.(let f = λb.a in f) in fst
```

The type for `fst` is  $\forall A \forall B. A \rightarrow B \rightarrow A$ .

Note that a naive typing could look like  $\forall A. A \rightarrow (\forall B. B \rightarrow A)$  but rank-2 polymorphism is not typable in Hindley-Milner. An important realization is the context from where the type analysis is made. If type analysis is made from within the bounded context of `f` the type of `f` becomes  $\forall B. B \rightarrow A$  and the type variable  $A$  is free.

The variables which may appear in a quantification have an important role in Equation 2.14, since only free variables may be substituted. Free variables are also a core part of generalizing a type for inference algorithms (subsection 2.2.1). When modelling polymorphic types with a technique such as Remark 2.1.1 finding the set of bound variables is trivial.

$$\text{bound}(\tau) = \text{free}(\tau) - \text{free}(\Gamma) \quad (2.11)$$

When generalizing a type  $\tau$  all types which do not occur in  $\Gamma$  must be quantified.

**Example 2.1.3.**

$$\Gamma = \{(x, \gamma)\} \quad (2.12)$$

$$\begin{aligned} \text{bound}(\tau \rightarrow \gamma) &= \{\tau, \gamma\} - \text{free}(\Gamma) \\ &= \{\tau, \gamma\} - \{\gamma\} = \{\tau\} \end{aligned} \quad (2.13)$$

Clearly the only bound type variable in the context of  $\tau \rightarrow \gamma$  is  $\tau$  such that it may become  $\forall \tau. \tau \rightarrow \gamma$  in the instance that the type represents a polymorphic let expression. Note that  $\mathbf{x}:\gamma$  in Equation 2.12 does not contain  $\gamma$  as a quantified type since it has been introduced by an abstraction and **Abs** only introduces monomorphic types (Figure 2.2). An interesting observation is that there can only exist one implementation of the above type system if  $\tau \rightarrow \gamma$  is to be introduced by a polymorphic let expression which is displayed in Listing 2.4 [Wad89].

Listing 2.4: Implementation of type state

```

1   $\lambda \mathbf{x}.$ 
2       $\text{let } \mathbf{z} = (\lambda \mathbf{y}. \mathbf{x}) \text{ in}$ 
3      ...

```

## 2.2 Hindley-Milner

With the now introduced primitives, the Hindley-Milner type system is but a set of rules composed by said primitives. There are six rules in the Hindley-

$$\begin{array}{c}
 \text{Var} \frac{x : \sigma \in \Gamma}{\Gamma \vdash x : \sigma} \\
 \\
 \text{App} \frac{\Gamma \vdash e_1 : \tau_1 \rightarrow \tau_2 \quad \Gamma \vdash e_2 : \tau_1}{\Gamma \vdash e_1 e_2 : \tau_2} \\
 \\
 \text{Abs} \frac{\Gamma, x : \tau_1 \vdash e : \tau_2}{\Gamma \vdash \lambda x. e : \tau_1 \rightarrow \tau_2} \\
 \\
 \text{Let} \frac{\Gamma \vdash e_1 : \sigma \quad \Gamma, x : \sigma \vdash e_2 : \tau}{\Gamma \vdash \text{let } x = e_1 \text{ in } e_2 : \tau} \\
 \\
 \text{Inst} \frac{\Gamma \vdash e : \sigma_1 \quad \sigma_1 \sqsubseteq \sigma_2}{\Gamma \vdash e : \sigma_2} \\
 \\
 \text{Gen} \frac{\Gamma \vdash e : \sigma \quad a \notin \text{free}(\Gamma)}{\Gamma \vdash e : \forall a. \sigma}
 \end{array}$$

Figure 2.2: Hindley-Milner type rules

Milner rules outlined in Figure 2.2.

- **Var** states that if some variable  $x$  with type  $\sigma$  exists in the environment, the type can be judged. In practice, when  $x : \sigma$  is encountered in the expression tree it is added to the environment.

- **App** decides that if  $e_1 : \tau_1 \rightarrow \tau_2$  and  $e_2 : \tau_1$  has been judged to exist then  $e_1 e_2$  implies the removal of  $\tau_1$  from  $\tau_1 \rightarrow \tau_2$  such that  $e_1 e_2 : \tau_2$ .
- **Abs** is the typing rule of lambda abstractions. If  $x : \tau_1$  exists in the environment from some type analysis of  $e$  and the abstraction's body  $e$  has been judged to be of type  $\tau_2$  then the abstraction of  $x$  must take the type of  $x$  to create the type of the body  $e$ .
- **Let** states that if  $e_1$  has been judged to have type  $\sigma$  then the let expression's identifier  $x : \sigma$  must exist in the environment when deriving the type of  $e_2$ . Observe that **Let** introduces a polymorphic type to the environment while **Abs** introduces a monomorphic one. Note that by Remark 2.1.2  $x$  may be polymorphic in  $e_2$ .
- **Inst** specializes some polymorphic type (in regard to the type system implementation) to a more specific polymorphic type.  $\sqsubseteq$  is the partial order of types where the binary relation between two types compares the descriptiveness of types.

**Example 2.2.1.** In  $L$  the smallest element is the top of the type hierarchy (Figure 2.1), the polymorphic type.

- **Gen** generalizes over all bound variables  $a$ .

Let polymorphism is exemplified in Example 2.2.2.

**Example 2.2.2.** Throughout this example the convenient syntax  $(x, z)$  is the pair of the variables  $x$  and  $z$  which can be implemented by algebraic data structures or a combinator.

The identity function is a common example to illustrate type systems (Listing 2.5).

Listing 2.5: Identity function in  $L$

```
1 fun id x = x;
2 id 4;
```

Listing 2.6: Identity function in lambda calculus with let

```
1 let id = ( $\lambda x.x$ ) in
2 id 4
```

Stating that `id` has the type  $\forall a.a \rightarrow a$  and `4` has the type `Int` is Listing 2.6 program correct? By applying the Hindley-Milner rules one can prove or disprove this statement. A correct proof of Listing 2.5 must be Figure 2.3.

Listing 2.7: Identity function in lambda calculus by abstraction

```
1 ( $\lambda id.id$  4) ( $\lambda x.x$ )
```

Listing 2.6 and Listing 2.7 are two equivalent programs with slightly different proofs which raises the question of why the `let` expression is even needed. If Listing 2.6 and Listing 2.7 were to be slightly changed such that two new programs Listing 2.8 and Listing 2.9 were to be proved, Listing 2.9 would not be provable while Listing 2.8 would.

Listing 2.8: Identity function with two applications

```
1 let id = (λx.x) in
2 (id 4, id id)
```

Listing 2.9: Identity function with two applications as abstraction

```
1 (λid.(id 4, id id)(λx.x))
```

In Listing 2.9 `id` cannot adhere to polymorphism by **Abs** in Figure 2.2 whilst **Let** can.

$$\begin{array}{c}
 \text{Var} \frac{id : \forall a.a \rightarrow a \in \Gamma}{\Gamma \vdash id : \forall a.a \rightarrow a} \quad \frac{\forall a.a \rightarrow a \sqsubseteq \text{Int} \rightarrow \text{Int}}{\Gamma \vdash id : \text{Int} \rightarrow \text{Int}} \quad \frac{4 : \text{Int} \in \Gamma}{\Gamma \vdash 4 : \text{Int}} \text{Var} \\
 \text{Inst} \frac{}{\Gamma, id : \forall a.a \rightarrow a \vdash id\ 4 : \text{Int}} \text{App}
 \end{array}
 \quad (a)$$

$$\begin{array}{c}
 \text{Var} \frac{x : a \in \Gamma}{\Gamma \vdash x : a} \\
 \text{Abs} \frac{}{\Gamma \vdash (\lambda x.x) : a \rightarrow a} \quad a \notin \text{free}(\Gamma) \\
 \text{Gen} \frac{}{\Gamma \vdash (\lambda x.x) : \forall a.a \rightarrow a}
 \end{array}
 \quad (b)$$

$$\text{Let} \frac{2.3b \quad 2.3a}{\Gamma \vdash \text{let id} = (\lambda x.x) \text{ in id 4 : Int}$$

Figure 2.3: Identity function instantiation proof

### 2.2.1 Damas-Milner Algorithm W

Typing rules are by themselves not that useful since they need all type information declared ahead of checking, inference attempts to guess types such that the rules are satisfied. Type inference is the technique of automatically deriving types, of which there exist many algorithms. One of the most common inference algorithms that produce typings which the Hindley-Milner rules accept is the Damas-Milner Algorithm W inference algorithm [Dam84; DM82].

The Damas-Milner Algorithm W rules (Figure 2.6) introduce some new concepts such as *fresh variables*, *most general unifier*, and the *substitution*

$S\Gamma = \{(x, S\sigma) \mid \forall(x, \sigma) \in \Gamma\}$	(Environment)
$S\sigma = \begin{cases} S\tau & \text{if } \sigma \equiv \tau \\ \{a' \mapsto \tau_1 \mid (a', \tau_1) \in S \mid (a, *) \notin S\}\sigma' & \text{if } \sigma \equiv \forall a. \sigma' \end{cases}$	(Poly)
$S(\tau_1 \rightarrow \tau_2) = S\tau_1 \rightarrow S\tau_2$	(Arrow)
$Sa = \begin{cases} \tau & \text{if } (a, \tau) \in S \\ a & \end{cases}$	(Typevariable)
$SC\tau_1 \dots \tau_n = CS\tau_1 \dots S\tau_n$	(Constructor)

Figure 2.4: Substitution semantics

*set*. Fresh variables are introduced by picking a variable that has not been picked before from the infinite set  $\tau_1, \tau_2 \dots$ . Fresh variables are introduced when unknown types are discovered and later unified. The substitution set is a mapping from type variables to types (Equation 2.14).

$$S = \{a_1 \mapsto \tau_1, a_2 \mapsto \tau_2 \dots, a_n \mapsto \tau_n\} \quad (2.14)$$

A substitution written  $ST$  where  $T$  is an arbitrary component of Hindley-Milner like an environment in which all type variables are substituted (Figure 2.4). Substitution sets can also be combined  $S_1 \cdot S_2$  with well defined semantics. The combination of substitution sets is a key component for the correctness of the Damas-Milner inference algorithm.

$$S_1 \cdot S_2 = \{(a \mapsto S_1\tau) \mid (a \mapsto \tau) \in S_2\} \cup S_1 \quad (2.15)$$

**Remark 2.2.1.** By the substitution set combination operator transitive and circular substitutions cannot occur since type variables in  $S_1$  will inherit all the mappings from  $S_2$  by union. Transitivity is avoided by substituting all instances of type variables values (the mapped to type variables) in  $S_2$  with ones that occur in  $S_1$ . The properties ensured by the combination semantics also induce the property of idempotence. This property is enforced by the Damas-Milner Algorithm W inference rules.

Unification is performed differently based on the context. Unification is performed on monotypes, each of which can take one of three forms (Equation 2.3). Note that the Var rules for most general unifier outlined in Figure 2.5 are commutative.

**Remark 2.2.2.** The Damas-Milner algorithm W is the most popular inference algorithm for Hindley-Milner. Though it remains the most popular, it has some interesting competitors. One of which is that of the constraint

$$\begin{array}{c}
\text{Arrow} \frac{S, \{(\tau_1 \rightarrow \tau_2, \gamma_1 \rightarrow \gamma_2)\} \cup T}{S, T \cup \{(\tau_1, \gamma_1), (\tau_2, \gamma_2)\}} \\
\\
\text{Intro} \frac{\tau_1, \tau_2}{\emptyset, \{(\tau_1, \tau_2)\}} \\
\\
\text{Var empty} \frac{S, \{(a, \tau_1)\} \cup T \quad a \equiv \tau_1}{S, T} \\
\\
\text{Var sub} \frac{S, \{(a, \tau_1)\} \cup T \quad a \notin \text{free}(\tau_1)}{S \cup \{a \mapsto S\tau_1\}, \{a \mapsto S\tau_1\} \cup T} \\
\\
\text{Atom} \frac{S, C_1\tau_1 \dots \tau_n, C_2\gamma_1 \dots \gamma_n \cup T \quad C_1 \equiv C_2}{S, \{(\tau_1, \gamma_1) \dots, (\tau_n, \gamma_n)\} \cup T}
\end{array}$$

Figure 2.5: Rules for most general unification

solver approach which is also used in OCaml [HHS02]. The constraint solver approach is a two phase type inference algorithm. In the first phase the algorithm inspects the expression tree and generates a set of constraints as it goes. After the set of constraints  $C$  has been generated it then traverses the constraints and generates type variable substitutions. It is argued that error reporting is significantly easier in such an approach.

**Remark 2.2.3.** In the language  $L$  a function **fun** translates to a let expressions while **let** translates to abstraction and application.

### 2.2.2 Instantiation

Another interesting addition introduced by algorithm W in Figure 2.6 is *inst*. *inst* naturally follows from the **Inst** rule in Figure 2.2 but has a slightly different behaviour. The *inst* function does not specify types anymore but simply makes unification of polymorphic types possible.

$$\text{inst}(\sigma) = \{a \mapsto \text{fresh} \mid a \notin \text{free}(\sigma)\} \sigma \quad (2.16)$$

*inst* (Equation 2.16) maps all bound type variables to fresh type variables in the polytype  $\sigma$ . *inst* is an important component to allow polymorphic types to remain polymorphic since no bound type variables may be substituted.

**Example 2.2.3.** Performing some type analysis on Listing 2.10 yields a very rich example of why *inst* is necessary.

$$\begin{array}{c}
\text{Var} \frac{x : \sigma \in \Gamma \quad \tau = \text{inst}(\sigma)}{\Gamma \vdash x : \tau, \emptyset} \\
\\
\text{Abs} \frac{\tau_1 = \text{fresh} \quad \Gamma, x : \tau_1 \vdash e : \tau_2, S}{\Gamma \vdash \lambda x. e : S\tau_1 \rightarrow \tau_2, S} \\
\\
\text{App} \\
\frac{\Gamma \vdash e_1 : \tau_1, S_1 \quad \tau_3 = \text{fresh} \quad S_1\Gamma \vdash e_2 : \tau_2, S_2 \quad S_3 = \text{mgu}(S_2\tau_1, \tau_2 \rightarrow \tau_3)}{\Gamma \vdash e_1 e_2 : S_3\tau_3, S_3 \cdot S_2 \cdot S_1} \\
\\
\text{Let} \frac{\Gamma \vdash e_1 : \tau_1, S_1 \quad S_1\Gamma, x : S_1\Gamma(\tau_1) \vdash e_2 : \tau_2, S_2}{\Gamma \vdash \text{let } x = e_1 \text{ in } e_2 : \tau_2, S_1 \cdot S_2}
\end{array}$$

Figure 2.6: Algorithm W

Listing 2.10: Polymorphic id

```

1 fun id x = x;
2 fun ap x f = f x;
3 fun doubleid x = id (id (x + 1));

```

After inferring `id` and `ap` the environment will contain `id`  $\Gamma = \{(\text{id}, \forall a. a \rightarrow a), (\text{ap}, \forall \gamma, \beta. \gamma \rightarrow (\gamma \rightarrow \beta) \rightarrow \beta)\}$ . Typing the function `doubleid` without the use of *inst*; begin by looking at the introduced parameter `x` and then the innermost expression `id (x + 1)`.

$$\text{bound}(\tau) = \text{free}(\tau) - \text{free}(\Gamma) = \{\tau\} \quad (\mathbf{Abs} \text{ intro } x : \tau)$$

$$\Gamma = \{(\text{id}, \forall a. a \rightarrow a), (\text{ap}, \forall \gamma, \beta. \gamma \rightarrow (\gamma \rightarrow \beta) \rightarrow \beta), (x, \forall \tau. \tau)\} \quad (2.17)$$

$$\text{unify}(\tau, \text{Int}) = \{\tau \mapsto \text{Int}\} \quad (2.18)$$

$$\text{unify}(a \rightarrow a, \text{Int} \rightarrow \mu) = \text{unify}(\{a \mapsto \text{Int}\}a, \{a \mapsto \text{Int}\}\mu) \cdot \{a \mapsto \text{Int}\} \quad (2.19)$$

$$= \{\mu \mapsto \text{Int}\} \cdot \{a \mapsto \text{Int}\}$$

$$= \{\mu \mapsto \text{Int}, a \mapsto \text{Int}\}$$

This example might not look compromising but a minor change such that the body of `doubleid` becomes `id (ap (id (x + 1)))` yields an interesting problem. In the case of introducing `ap` the two type instances for `id` must be different (`id` must be introduced with different type variables) to retain its polymorphic properties. The following steps are performed when inferring

this new body.

$$\begin{aligned}
& \text{unify}(\gamma \rightarrow (\gamma \rightarrow \beta) \rightarrow \beta, \text{Int} \rightarrow \delta) && (\text{ap } (\text{id } (\mathbf{x} + 1))) \\
& = \{\gamma \mapsto \text{Int}, \delta \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\} \\
& \{\gamma \mapsto \text{Int}, \delta \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\} \cdot \{\mu \mapsto \text{Int}, a \mapsto \text{Int}\} && (\mathbf{App } S_3 \cdot S_2) \\
& = \{\gamma \mapsto \text{Int}, \delta \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta, \mu \mapsto \text{Int}, a \mapsto \text{Int}\} \\
& \text{unify}(a \rightarrow a, ((\text{Int} \rightarrow \beta) \rightarrow \beta) \rightarrow \theta) && (\text{id } (\text{ap } (\text{id } (\mathbf{x} + 1)))) \\
& = \text{unify}(\{a \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\}a, \{a \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\}\theta) \cdot \{a \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\} \\
& = \{a \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta, \theta \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\} \cdot \{a \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\} \\
& = \{a \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta, \theta \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\} \\
& \{a \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta, \theta \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta\} \cdot && (\mathbf{App } S_3 \cdot S_2) \\
& \{\gamma \mapsto \text{Int}, \delta \mapsto (\text{Int} \rightarrow \beta) \rightarrow \beta, \mu \mapsto \text{Int}, a \mapsto \text{Int}\}
\end{aligned}$$

Clearly  $a$  cannot map to two types which cannot be unified which is a violation of the type system. The apparent problem is that `id` is specialized within the whole of `doubleid`. By instantiating quantified types when they are needed cases such as this can be avoided (it also makes the algorithm correct).

$$\begin{aligned}
& \text{unify}(\text{inst}(\forall a. a \rightarrow a), \text{inst}(\forall \tau. \tau \rightarrow \mu)) && (2.20) \\
& = \text{unify}(\gamma \rightarrow \gamma, \varphi \rightarrow \mu) \\
& = \{\varphi \mapsto \mu, \gamma \mapsto \mu\}
\end{aligned}$$

### 2.2.3 Recursion

Recursion is a trivial matter once the primitives of the Hindley-Milner type system have been introduced. Recall that in subsection 1.2.2 recursion (along with mutual recursion) was shown to be implementable by introducing functions to their own scope, the same is true for types. Allowing recursive functions in Hindley-Milner type inference systems is a matter of letting the function be present in the environment when inferring the function's own body.

**Example 2.2.4.** If the function `f` defined in Listing 2.11 were to be typed it would need to be introduced as an unknown type to the environment before typing the body of `f`.

Listing 2.11: Recursive function

```
1 fun f x = (f x) + 1;
```

Let  $\Gamma = \{f : \forall \tau. \tau, x : \forall \mu. \mu\}$ . From the application `f x` the unification  $\text{unify}(\tau, \gamma \rightarrow \mu) = \{\tau \mapsto \mu \rightarrow \gamma\}$  must be performed, and the resulting type for the expression is  $\gamma$ . The addition operation forces  $\text{unify}(\text{Int}, \gamma) = \{\gamma \mapsto$



`Int`}. Finally the application of the addition function  $+$  : `Int`  $\rightarrow$  `Int`  $\rightarrow$  `Int` and the two expressions `f x` and `1` such that the resulting expression type is `Int`.

## 2.2.4 Additional language features

In addition to the rules in Figure 2.6 many other ergonomic features can easily be modelled once the framework has been understood. One of the most crucial features of languages are that of decision.

Listing 2.12: ADT implementation of decision

```

1 type Boolean =
2   | BFalse
3   | BTrue
4 ;
5 let b = BFalse;
6 fun double x = x + x;
7 b (0) (double 10);

```

Decision can be implemented in a variety of ways such as in Listing 2.12 by the use of algebraic data structures aligning very much with Church Booleans [Chu85]. Rather decision can be implemented by more conven-

$$\begin{array}{c}
\Gamma \vdash e_2 : \tau_2, S_2 \quad \tau_4 = \text{fresh} \quad \Gamma, S_4 = \text{mgu}(S'_1 \cdot S_2 \cdot S_3 \tau_2, \tau_4) \\
\\
\text{(a)} \\
\\
\Gamma \vdash e_1 : \tau_1, S_1 \quad S'_1 = \text{mgu}(\text{Bool}, \tau_1) \quad \Gamma \vdash e_3 : \tau_3, S_3 \quad \Gamma, S_5 = \text{mgu}(S_4 \cdot S'_1 \cdot S_2 \cdot S_3 \tau_3, S_4 \tau_4) \\
\\
\text{(b)} \qquad \qquad \qquad \text{(c)} \\
\\
\frac{\text{2.7b} \quad \frac{\text{2.7a} \quad \text{2.7c}}{\Gamma, S_6 \quad S_6 = S_5 \cdot S_4 \cdot S'_1 \cdot S_2 \cdot S_3}}{\Gamma \vdash \text{if } e_1 \text{ then } e_2 \text{ else } e_3 : S_6 \tau_4, S_6}
\end{array}$$

Figure 2.7: Decision

tional methods than combinator logic by introducing more inference rules as in Figure 2.7. Additional language syntax features can in most cases be implemented as decision can.

## 2.2.5 Algebraic data structures

To implement rules for algebraic data structures one must first decide on what the type of an algebraic data structure is. If algebraic data struc-

tures were implemented as in subsection 1.3.1 the type of an algebraic data structure like `Boolean` in Listing 2.12 would be  $a \rightarrow a \rightarrow a$  since `BFalse` and `BTrue` must have a handler each. Implementing algebraic data structures by this method does not introduce anything to the inference algorithm since every algebraic data structure becomes a function. It is more common to introduce algebraic data structures as new type constructor types since such an implementation yields descriptive errors in comparison to generated function types. Fortunately type constructors are trivial to model in Hindley-Milner since a type constructor is a type lambda and rank-1 type lambdas are simply quantified types. Listing 1.17 introduces the type constructor  $a \rightarrow \text{List } a$  with value constructors `Cons`:  $\forall a. a \rightarrow \text{List } a \rightarrow \text{List } a$  and `Nil`:  $\forall a. \text{List } a$ . For instance let some program be `Cons 1 (Cons 2 Nil)` such that inference becomes a matter of first unifying the type of `Cons` and  $\forall a. \text{Int} \rightarrow \text{List } a \rightarrow \tau_1$  using the application rule and then unifying  $\forall a. \text{Int} \rightarrow \tau_1 \rightarrow \tau_2$  with `Cons` again by application.

## 2.3 The cost of expressiveness

Modern languages with strong type systems tend to be notoriously slow to type on pathological inputs. In fact, many languages with strong type systems provide type systems expressive enough to be Turing-complete.

In the construction of the compiler for *L*, one target was the C++ language. An instance of a pathological input for the C++ type checker is most definitely the untyped lambda calculus. The lambda terms in C++ must adhere to polymorphism in many cases which leads to some unknown but large blowup in compilation time. In fact type polymorphism is commonly the root of blowup in typing.

ML, which implements a Hindley-Milner inference system, was believed to have linear complexity before shown to be exponential along with other problematic complexity findings [Mai89]. As it will turn out, Hindley-Milner also suffers an explosive worst case induced by a pathological input fueled by polymorphism.

**Lemma 2.3.1.** *There exists a family of programs which are typable in Hindley-Milner and produce  $\Omega(2^n)$  unique type variables.*

*Proof.* The basis of the blowup stems from the introduced fresh type variables in the polymorphic `Let` inference rule. If the number of type variables can be shown to be exponential, the running time must be at least the same by operations, such as subsection set combination and unification.

Listing 2.13: Nested dup

```

1 fun dup a f = f a a;
2 fun deep x = dup (dup (dup (... x)));

```

Listing 2.13 builds a large function signature for **deep**. The innermost **dup** invocation will have its signature unified to  $x \rightarrow (x \rightarrow x \rightarrow \tau) \rightarrow \tau$ , if **a** has type  $x$  and **f** has type  $x \rightarrow x \rightarrow \tau$  for some unknown  $\tau$  by the App rule in Figure 2.6. The second innermost **dup** invocation has the signature  $((x \rightarrow x \rightarrow \tau) \rightarrow \tau) \rightarrow (((x \rightarrow x \rightarrow \tau) \rightarrow \tau) \rightarrow \gamma) \rightarrow \gamma$ . Naively one might judge Listing 2.13 to run in  $\Omega(2^n)$  but an important observation for why Listing 2.13 does not induce exponential blowup is the uniqueness of the type variables. If an efficient representation of **dup** was implemented such that the left and right side were shared such that  $\mu \mapsto ((x \rightarrow x \rightarrow \tau) \rightarrow \tau)$ , the number introduced type variables would be  $O(n)$ .

Listing 2.14: Nested tuples with different type variables

```

1 fun tuple a b f = f a b;
2 fun one = tuple tuple tuple;
3 fun two = tuple one one;
4 fun three = tuple two two;
5 ...

```

The trick to induce an exponential running time is demonstrated with the pathological program in Listing 2.14. By allowing **tuple** to be polymorphic and have having two polymorphic parameters, every time **tuple** is instantiated, it will contain only fresh variables. The type of **tuple** is  $a \rightarrow b \rightarrow (a \rightarrow b \rightarrow c) \rightarrow c$ . Clearly this looks very much like Listing 2.13, but has the subtle difference of letting the parameters **a** and **b** (within the type instantiation of the let expression **tuple**) be polymorphic and introducing every "step" as a polymorphic let expression. The return type of **one** (the type of **f**) is displayed in Equation 2.21.

$$\text{inst}(a \rightarrow b \rightarrow (a \rightarrow b \rightarrow c) \rightarrow c) \rightarrow \text{inst}(a \rightarrow b \rightarrow (a \rightarrow b \rightarrow c) \rightarrow c) \rightarrow \gamma \rightarrow \gamma. \quad (2.21)$$

The first and second instantiations will contain different type variable such that they are not structurally equivalent (Equation 2.22).

$$(\tau \rightarrow \mu \rightarrow (\tau \rightarrow \mu \rightarrow \phi) \rightarrow \phi) \rightarrow (\varphi \rightarrow \zeta \rightarrow (\varphi \rightarrow \zeta \rightarrow \delta) \rightarrow \delta) \rightarrow \gamma \rightarrow \gamma. \quad (2.22)$$

□

An interesting observation is that by increasing the amount of polymorphic parameters to some  $c$  the number of type variables becomes  $\Omega(c^n)$ . This observation does not have any significant impact since  $O(f(n)) \supseteq \Omega(n^n)$  where  $f$  is the algorithm for type inference, such that the problem of type inference in Hindley-Milner is at least in EXPTIME which solvable in both  $O(2^n)$  and  $O(n^n)$ . The upper bound which states that type inference in Hindley-Milner is in fact EXPTIME-complete was justified in [KTU90;

Mai89]. Running the program Listing 2.14 in  $L$  yields a blowup of  $2^n$  (Listing 5.1). Figure 2.8 shows the relationship between the program typed in  $L$  and the theoretical time of  $2^n$ .

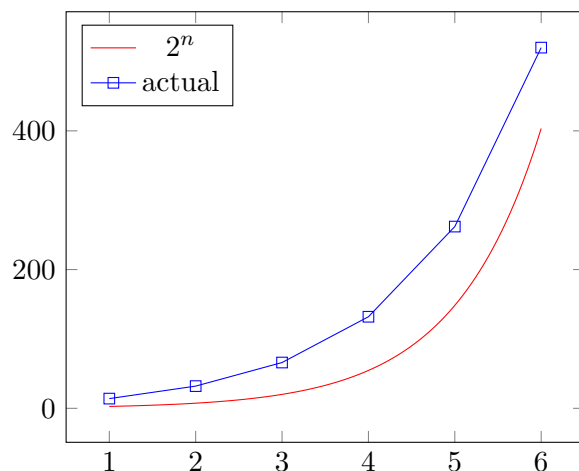


Figure 2.8: Plot of type variables in Hindley-Milner type systems

## 2.4 Higher level type systems

The Hindley-Milner type system can only express relatively simple programs which robs algorithmic elegance in respect to other type systems. One domain of programs that Hindley-Milner cannot express are those that rely on *rank-n types*. Rank-n types deals with letting abstractions have polymorphic parameters such that a type can be quantified within another type, having its depth bounded by  $n$  (rank- $n$ ). For instance Listing 2.15 is not typable in Hindley-Milner since its type is  $\forall \tau. (\forall \gamma. \gamma \rightarrow \text{Int}) \rightarrow \tau \rightarrow \text{Int}$ .

Listing 2.15: Program that requires rank-n types

```

1 fun f makeNum a =
2   ((makeNum a) + (makeNum 0)) + (makeNum (0 == 2))

```

More generally, any type which is quantified on the left side of  $\rightarrow$  cannot be moved out thus increases the rank.

Even languages which are typed and inferred by Hindley-Milner like Ocaml have introduced kinds through modules to allow higher-kinded types. Hindley-Milner is in fact a restricted version of another more general type system called *System F* (and *System F $\omega$* ). The Hindley-Milner type system introduces abstractions as monomorphic types whereas System F allows any type to be polymorphic. It turns out that allowing higher rank polymorphism makes type inference (type reconstruction in older literature) *undecidable* [Wel99].

**Remark 2.4.1.** Formal type systems are in their essence deductive systems, which have provable properties such as *decidability*. Decidability in deductive systems is a property which expresses whether a system can be decided by an algorithm (which relates to the encoding of algorithms on theoretic computers). If and only if every valid formula (type) in the deductive system (type system) can have its correctness decided (and reconstructed if necessary) algorithmically.

Another variant of type system is *System  $F_{\omega}$* . System  $F_{\omega}$  introduces another feature (System  $F_{\omega}$  is different to System F, it is not an extension) called type constructors. It is uncommon to use System  $F_{\omega}$  on its own since it only allows type constructors of monomorphic types (System F introduces polymorphism), which does not yield much expressiveness since only specific types such as  $\text{Int} \rightarrow \text{List Int}$  would be expressible. Throughout this chapter, type constructors have already been introduced in such a way that they can occur in Hindley-Milner though algebraic data types such as  $\forall a. a \rightarrow \text{List } a$ . Very commonly, moderately generalized types need both the higher rank polymorphism implied by System F and the type constructors implied by System  $F_{\omega}$ .

Hindley-Milner can only take advantage of System  $F_{\omega}$  for rank 1 types which significantly constrains the generalization level. A more expressive version of Hindley-Milner is System  $F_{\omega}$  which in fact, is the basis for the type system of Haskell, which is significantly more expressive than Hindley-Milner.

**Remark 2.4.2.** Haskell has introduced some additional tweaks to System  $F_{\omega}$  to avoid the decidability problem among others.

In more expressive functional programming language type systems it has become increasingly popular abstract over implementations by introducing concepts from *category theory*. Naturally many abstractions of category theory require rank-2 polymorphism. More generally the larger the level of polymorphism allowed the larger the possible abstraction level becomes. For instance a general purpose *functor* is implementable and usable with rank-2 polymorphism while a natural transformation becomes a matter of rank-3 polymorphism.

**Remark 2.4.3.** A functor is a mapping that maps from type constructor instance to another, which for instance can be a functor for lists which provides the algebra  $\forall a. \forall b. \text{List } a \rightarrow \text{List } b$ .

To generalize functor one must be able to express *kinds* which are the types of type constructors denoted  $* \rightarrow *$  for a type constructor that takes some type  $*$  and creates some type  $*$ .  $* \rightarrow *$  is a unary type constructor whereas  $*$  is an atomic type like  $\text{Int}$  or  $\text{List Int}$ , since these types are fully applied. Kinds allow partial application on type constructors on a general level, since

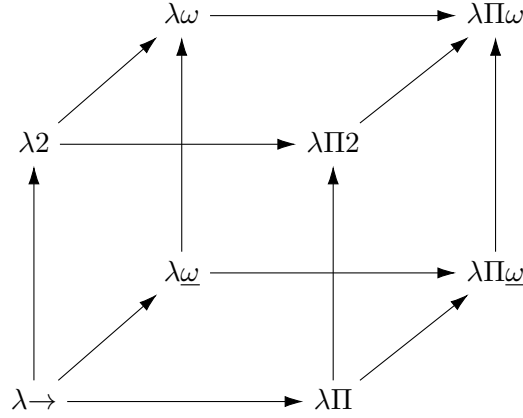


Figure 2.9:

- $\lambda \rightarrow$  is the simply typed lambda calculus without polymorphism.
- $\lambda \underline{\omega}$  is System  $F\underline{\omega}$ .
- $\lambda 2$  is System  $F$ .
- $\lambda \omega$  is System  $F\omega$ .
- $\Pi$  introduces *dependent types* which is beyond the scope of this thesis.

the only specific constraint is the shape and not where variables appear. The relaxation of type constructions allow various types to be generalized such as  $\mathbf{a} \rightarrow \mathbf{b} \rightarrow \mathbf{M} \ \mathbf{a} \ \mathbf{b}$  which could also have the signature of  $\mathbf{a} \rightarrow \mathbf{b} \rightarrow \mathbf{M} \ \mathbf{b} \ \mathbf{a}$  which kinds abstract over generalizing  $M$  to  $* \rightarrow * \rightarrow *$ . The kind for `List` is  $* \rightarrow *$  such that for any  $\tau$  with kind  $* \rightarrow *$  the type for functor map is  $\forall \tau. (\forall \mathbf{a}. \tau \mathbf{a}) \rightarrow (\forall \mathbf{b}. \tau \mathbf{b})$ .

**Remark 2.4.4.** Kinds are an abstraction which can exist purely theoretical without robbing the type system of expressiveness. Just as some complications in type systems are resolved with weakening the type system or enriching the syntax, kinds can be abstracted away into types [WHE13].

$\lambda P$  introduces *Dependent types* which lets types depend on terms in the language. A common example to show what dependent types can do is that of combining two lists  $v_1$  and  $v_2$  of size  $n_1$  and  $n_2$  into a list  $v_3$  of size  $n_1 + n_2$ , where the size can be expressed in the type system. The signature for such a function in  $L$  could be  $\forall a. \text{List } n_1 \ \mathbf{a} \rightarrow \text{List } n_2 \ \mathbf{a} \rightarrow \text{List } (n_1 + n_2) \ \mathbf{a}$ . Clearly one would have rules for lists such as a base case for the empty list `fun empty = Nil` with type  $\forall a. \text{List } 0 \ \mathbf{a}$ , which indicates how the type system is lifted to a logical proofing tool.

Figure 2.9 shows the *lambda cube*, introduced in [Bar91] which encapsulates the family of formal type systems. Complicated type system such as the *calculus of constructions* ( $\lambda\Pi\omega$ ) are used in proof assistants since they essentially are deduction systems.

## 2.5 Concluding remarks

This section should act as an introduction to more general type systems and where Hindley-Milner is placed on the type system map. Hindley-Milner is a small part of a larger more general system which has significant impact on the extensibility of Hindley-Milner. Some very renown functional programming languages began by implementing Hindley-Milner as their type system since it is very fast in practice and relatively simple to implement.

## Chapter 3

# Program evaluation

The untyped lambda calculus may provide a simple interface for programming but does not pair very well with the modern computer. *Interpreting* is a common technique for evaluating the untyped lambda calculus. An interpreter is an execution engine usually implemented in a more low-level language.

### 3.1 Evaluation strategies

When evaluating the untyped lambda calculus one has to choose an evaluation strategy. The choice of evaluation strategy has a large impact on aspects such as complexity guarantees. Such strategies are *call by value*, *call by name* and *call by need*. Call by value is most often the simplest and most natural way of assuming program execution.

Listing 3.1: Program that doubles values

```
1 fun double x = x + x;  
2 let a = double 10;  
3 double 10;
```

By the call by value semantics, Listing 3.1 eagerly evaluates every expression. Clearly the variable `a` is never used but under the call by value semantics everything is eagerly evaluated. Every expression is evaluated in logical order in the call by value evaluation strategy.



Listing 3.2: Implementation of call by name

```

1 fun suspend x unit = x;
2 fun force x = x 0;
3 let value = suspend 10;
4 fun double x =
5     fun susExpensiveOp unit =
6         (force x) + (force x);
7     susExpensiveOp;
8 let a = double value;
9 force (double value);

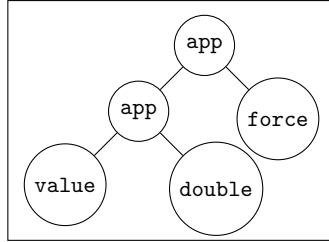
```

The call by name semantics however does only evaluate expressions once they are needed. By the call by name semantics `a` is never evaluated since it is never used. In Listing 3.2 call by name has been implemented by the use of various functions such as the two constant functions `suspend` and `force`. `susExpensiveOp` ensures that the forcing (evaluation) of `x` never occurs until the caller of `double` forces the result. By the aforementioned semantics of call by name in the context of the program in Listing 3.2 `a` is never forced thus the computation is never performed. The implementation of call by name can become quite troublesome and therefore in most cases is a part of the native execution environment which will be discussed in ??.

The call by need strategy introduces *lazy evaluation* semantics which is the same as call by name with one extra detail named *sharing*. In Listing 3.2 `force x` is computed twice which may be an expensive operation. Under call by need all results are saved for later use similar to techniques such as dynamic programming. To understand this better observe the expression tree for Listing 3.2 in Figure 3.1. Clearly the two red subtrees in Figure 3.1b are identical thus they may be memoized such that the forcing of `x` only occurs once. More generally if the execution environment supports lazy evaluation, once an expression has been forced it is remembered.

## 3.2 Runtime environments

Now that the untyped lambda calculus has been introduced, implemented and validated efficiently the question of execution naturally follows. There exists many different well understood strategies to implement an execution environment for the untyped lambda calculus. Naively it may seem straightforward to evaluate the untyped lambda calculus mechanically by  $\beta$ -reductions, but doing so brings upon some problems when implementing an interpreter.



(a) The last expression of the program.



(b) The expression tree for **double**

Figure 3.1

### 3.2.1 Combinator reducers

One of the most prominent techniques for evaluating functional programs is that of *combinator graphs reductions*. Formally a combinator is a function that has no free variables which is convenient since the problem of figuring out closures and parameter substitutions in applications never arises.

$$x \quad (3.1)$$

$$F \quad (3.2)$$

$$YE \quad (3.3)$$

There are three types of terms in combinator logic; the variable much like the lambda calculus (Equation 3.1), application (Equation 3.3) and the combinator (Equation 3.2). The SKI calculus is a very simple set of combinators which are powerful enough to be turing complete and translate to and from the lambda calculus. In SKI  $F ::= S \mid K \mid I$  where the equivalent lambda calculus combinators for  $S = \lambda x.\lambda y.\lambda z.xz(yz)$ ,  $K = \lambda x.\lambda y.x$  and  $I = \lambda x.x$ . Evaluating an SKI program is a straightforward reduction where  $F'_F$  denotes combinator  $F'$  has been partially applied with combinator  $F$ .

**Example 3.2.1.**

$$\begin{aligned} & SKSI \\ &= KI(SI) \\ &= K_I(SI) \\ &= I \end{aligned} \quad (3.4)$$

The algorithm for converting a lambda calculus program into a SKI combinator program is a straightforward mechanical one. The evaluation context is always an abstraction  $\lambda x.E$ .

Case 1:  $E = x$  then rewrite  $\lambda x.E$  to  $I$ .

Case 2:  $E = y$  where  $y \neq x$  and  $y$  is a variable then rewrite  $\lambda x.y$  to  $Ky$ .

Case 3:  $E = YE'$  then rewrite  $\lambda x.YE'$  to  $S(\lambda x.Y)(\lambda x.E')$  since applying some  $y$  to  $\lambda x.YE'$  must lambda lift  $y$  as a parameter named  $x$  to both  $Y$  and  $E'$  such that the lifted expression becomes  $((\lambda x.Y)y)((\lambda x.E')y) = S(\lambda x.Y)(\lambda x.E')y$ . Then recurse in both branches.

Case 4:  $E = \lambda x.E'$  then first rewrite  $E'$  with the appropriate cases recursively such that  $E'$  becomes either  $x$ ,  $y$  or  $YE$  such that Case 1, 2 or 3 can be applied.

The termination of the rewriting to SKI is guaranteed since abstractions are always eliminated and the algorithm never introduce any additional abstractions. When translating the untyped lambda calculus to SKI the "magic" variable names  $\sigma, \kappa$  and  $\iota$  are used as placeholder functions for the SKI combinators since the translation requires a lambda calculus form. When the translation has been completed then replace  $\sigma \mapsto S, \kappa \mapsto K, \iota \mapsto I$ .

### 3.2.2 Combinator translation growth

Before proving that the SKI translation algorithm produces a program of larger size the notion of size must be established. Size in terms of lambda calculus are the amount of lambda terms (Equation 1.2, Equation 1.1 and Equation 1.3) that make up a program. For instance  $\lambda x.x$  has a size of two since it is composed of an abstraction and a variable term. The size of an SKI combinator program is in terms of the number of combinators.

**Lemma 3.2.1.** *There exists a family of lambda calculus programs of size  $n$  which are translated into SKI-expressions of size  $\Omega(n^2)$ .*

*Proof.*

Case 1: Rewriting  $\lambda x.x$  to  $I$  is a reduction of one.

Case 2: Rewriting  $\lambda x.y$  to  $Ky$  is equivalent in terms of size.

Case 3: Rewriting  $\lambda x.YE$  to  $S(\lambda x.Y)(\lambda x.E)$  is the interesting case. To induce the worst case size Case 1 must be avoided. If  $x \notin \text{Free}(Y)$  and  $x \notin \text{Free}(E)$  then for every non-recursive term in  $Y$  and  $E$  Case 2 is the only applicable rewrite rule which means that an at least equal size is guaranteed. Furthermore observe that by introducing unused parameters one can add one  $K$  term to *every* non-recursive case. Observe the instance  $\lambda f_1.\lambda f_2.\lambda f_3.(f_1 f_1 f_1)$  where the two unused parameters are used to add  $K$  terms to all non-recursive cases in Equation 3.5 such that the amount of extra  $K$  terms minus the  $I$  becomes  $\text{variable\_references} * (\text{unused\_abstractions} - 1) = 3 * (3 - 1)$ :

$$S(S(KKI)(KKI))(KKI) \quad (3.5)$$

Now let the number of variable references be  $n$  and the unused abstractions also be  $n$  clearly  $\Omega(n * (n - 1)) = \Omega(n^2)$

Case 4: Rewriting  $\lambda x.E'$  is not a translation rule so the cost is based on what  $E'$  becomes.

Notice that the applications  $f_1 f_1 \dots f_1$  can in fact be changed to  $f_1 f_2 \dots f_n$  since for every  $f_k$  where  $0 < k \leq n$  there are  $n - 1$  parameters that induce a  $K$  combinator. Let  $P_n$  be family of programs with  $n$  abstractions and  $n$  applications.  $\lambda f_1.\lambda f_2.\lambda f_3.(f_1 f_1 f_1) \in P_3$  and in fact for any  $p$  where  $\forall n \in \mathbb{Z}^+$  and  $p \in P_n$ ,  $p$  translates into SKI-expressions of size  $\Omega(n^2)$ .  $\square$

**Example 3.2.2.** Observe the size of Equation 3.6 in comparison to Equation 5.1.

$$\begin{aligned} & \lambda f_1.\lambda f_2.f_1 f_2 & (3.6) \\ = & \lambda f_1.\sigma(\lambda f_2.f_1)(\lambda f_2.f_2) \\ = & \lambda f_1.(\sigma(\kappa f_1))(\iota) \\ = & \sigma(\lambda f_1.\sigma(\kappa f_1))(\lambda f_1.\iota) \\ = & \sigma(\sigma(\lambda f_1.\sigma)(\lambda f_1.\kappa f_1))(\kappa \iota) \\ = & \sigma(\sigma(\kappa \sigma)(\sigma(\lambda f_1.\kappa)(\lambda f_1.f_1)))(\kappa \iota) \\ = & \sigma(\sigma(\kappa \sigma)(\sigma(\kappa \kappa)(\iota)))(\kappa \iota) \\ = & S(S(KS)(S(KK)(I)))(KI) \end{aligned}$$

It should become clear that many programs suffer from this consequence such as `let add = (λx.λy.(+ x) y) ∈ P2` where the program is written in prefix notation. Translating the lambda calculus into the SKI-expressions does indeed increase the size significantly but does not warrant a write off entirely. More advanced techniques exist to translate the lambda calculus to linearly sized SKI-expressions with the introduction of more complicated combinators [Kis18].

### 3.2.3 Reduction strategies

Reductions in the context of the lambda calculus are a small set of well-defined rules for rewriting such that a program is evaluated. The techniques required to correctly evaluate a program are a bit more complicated than the SKI calculus but are rewarding in flexibility and performance. The substitution mapping set, denoted  $S$ , is a set of variable names to their value denoted  $\{x \mapsto \lambda y.y\}$  meaning “the value of variable  $x$  is  $\lambda y.y$ ”.

$$\begin{aligned} \{x \mapsto y\}x &= y \\ \{x \mapsto y\}z &= z \end{aligned} \tag{3.7}$$

Substitution mappings are used as in Equation 3.8; if  $Sx = y$  if  $(x \mapsto y) \in S$  else  $Sx = x$ .

**Remark 3.2.1.** Let expressions are considered as if abstractions and applications; `let a = b in c`  $\rightarrow (\lambda a.c)b$ .

Evaluation strategies (section 3.1) are a core part of the reduction strategy since the choice of evaluation strategy changes the order in which terms are evaluated. A *redex* is a reducible expression in the context of some set of reduction rules. The two interesting evaluation orders are *applicative order* and *normal order* [Ses02]. Applicative order specifies that the parameters of some application should always be evaluated first, e.g. call by value. Normal order specifies that the leftmost outermost term should be evaluated first which yields the call by name strategy. Before delving into more complicated evaluation strategies such as call by need, call by name will be considered.

A naive reduction strategy would involve adding variables to the substitution set, once they are applied. When evaluating a term such as `(λx.x + 5) 5`,  $x$  must be substituted by `5` such that the expression becomes `x + 5` with the substitution mapping  $\{x \mapsto 5\}$ . The rules in Figure 3.2 display a simple set of rules for evaluating the call by name lambda calculus. The Abs and Var rules (Figure 3.2a and Figure 3.2b) are rules which act as leaves within the evaluation of a program. Abs and Var both state that if either of them occur then simply return the expression. The App rule (Figure 3.2c) states that “First evaluate  $f$  to  $(\lambda x.e)$ , then evaluate  $e$  to  $y$ , where  $x$  has

$$\begin{array}{ccc}
\frac{}{(\lambda x.e) \rightarrow (\lambda x.e)} \text{ Abs} & & \frac{}{x \rightarrow y} \text{ Var} \\
(a) & & (b) \\
\frac{f \rightarrow (\lambda x.e) \quad \{x \mapsto z\} e \rightarrow y}{f z \rightarrow y} \text{ App} & & \\
(c) \text{ A simple application rule} & & 
\end{array}$$

Figure 3.2: Simple call by name lambda calculus

been replaced with  $z$  in  $e$ ". We must introduce rules for how substitutions should act upon encountering lambda calculus term types (Equation 3.8).

$$\{x \mapsto y\} x = y \quad (3.8)$$

$$\{x \mapsto y\} z = z$$

$$S f z = (Sf)(Sz)$$

$$S (\lambda x.e) = (\lambda x.Se) \quad (3.9)$$

Listing 3.3: Program with variable ambiguity

```

1 fun f x =
2   fun g x = x;
3   g (x + x);
4 f 2;

```

Evaluating Listing 3.3 under the rules in Equation 3.8 yields a case for more thorough rules. Using the App rule (Figure 3.2c) and applying substitutions as defined in (Equation 3.8), yields the program in Listing 3.4.

Listing 3.4: Program with variable ambiguity after one reduction pass

```

1 fun g x = 2;
2 g (2 + 2);

```

Which eventually evaluates to 2, which clearly is not the intended result. Removing the rule Equation 3.9 and adding the two rules in Equation 3.10 solves variable ambiguity.

$$S(\lambda x.e) = (\lambda x.Se) \quad (x \mapsto y) \notin S \quad (3.10)$$

$$S(\lambda x.e) = (\lambda x.(S \setminus \{x \mapsto y\})e) \quad (x \mapsto y) \in S$$

This evaluation model is indeed powerful enough to evaluate the call by name lambda calculus naively.

To further refine the call by name evaluation semantics observe that substitutions are performed eagerly, that is, they *break suspension*. To explore solutions to avoid eager substitutions, environments may be used.

**Remark 3.2.2.** When considering an arbitrarily large expression under call by name (or need), there is no guarantee that the expression will ever be evaluated, that is, a suspended expression being forced. Breaking suspension violates the philosophy behind call by name (and need). A solution could involve composing a function in compile time, which replaces all instances of a variable (by Equation 3.8 and Equation 3.10) with some parameterized value. Substituting expressions eagerly might compromise performance guarantees, even if substitution is an asymptotically constant number of operations. Consider substituting an expression which contains conditional branches, one branch might be a single expression and the other might be a very large expression, consisting of many instances of the same variable. If the program never picks the large expression, why should it ever have any impact on performance. For instance, in Listing 3.5 under call by name one would expect `fastMap` to evaluate in roughly the same time as `slowMap`.

Listing 3.5: Performance considerations in substitution

```

1 fun fastMap f l =
2   match l
3     | Nil -> Nil;
4     | Cons x xs -> Cons (f x) (map f xs);
5   ;
6 fun slowMap f l =
7   match l
8     | Nil -> Nil;
9     | Cons x xs ->
10      if (x == 0)
11        let v = x + x + x + x + x + x;
12          Cons (f x) (map f xs);
13      else
14        Cons (f x) (map f xs);
15      ;
16 ;

```

## Environments

Environments, like environments in typing (Equation 2.5), define what “state” is required to evaluate some expression. In the case of reduction strategies, the environment is always a substitution mapping. Environments are a requirement for some call by need semantics and lazy substitution. A simple modification to the rules in Figure 3.3, implies an interpreter which brings

an environment along when interpreting. The  $\cup$  operator merges two sub-

$$\begin{array}{c}
 \frac{}{S, (\lambda x. e) \rightarrow S, (\lambda x. e)} \text{Abs} \\
 \text{(a)} \\
 \frac{\Theta, f \rightarrow S, (\lambda x. e) \quad S \cup \{x \mapsto z\}, e \rightarrow \Sigma, y}{\Theta, f z \rightarrow \Sigma, y} \text{App} \quad \frac{}{S, x \rightarrow S, Sx} \text{Var} \\
 \text{(b)} \qquad \qquad \qquad \text{(c)}
 \end{array}$$

Figure 3.3: Call by name lambda calculus with environments

stitution mappings, choosing the rightmost mapping on duplicates, such that  $\{x \mapsto y, z \mapsto h\} \cup \{x \mapsto m\} \rightarrow \{x \mapsto m, z \mapsto h\}$ , such that the semantics of  $\cup$  satisfy Equation 3.10.

Listing 3.6: Problematic program

```

1 fun f a =
2   fun g x =
3     let a = 20;
4     a + x;
5     (g a) + a;
6 f 10;

```

Evaluating Listing 3.6, using the rules proposed in Figure 3.3, yields an interesting case of variable ambiguity. At some point the evaluation machine will reach a state with the expression  $(g\ a) + a$  and substitution mapping  $\{a \mapsto 10, g \mapsto (\lambda x. \text{let } a = 20 \text{ in } (a + x))\}$ . By the laws of addition, that is, commutativity, it should be possible pick either of the two expressions to evaluate first. If any one of the branches in the addition operator are side effectful, it may become a problem, but this language only deals with pure programs. Evaluating the left side of  $(g\ a) + a$  first, yields  $30 + a$  with the substitution mapping  $\{a \mapsto 20, g \mapsto \dots\}$ , and finally 50. Evaluating the right side of  $(g\ a) + a$  first, yields  $(g\ a) + 10$  with the substitution mapping  $\{a \mapsto 10, g \mapsto \dots\}$ , and finally 40. Clearly the order of evaluation is of importance for the result, which is often not the desired semantics. If there exists multiple reduction techniques which ensure the same result, the system is called *confluent*. The reduction rules for the lambda calculus are confluent [CR36], which implies that the current set



of rules are not valid evaluation rules for the lambda calculus. If the rules were to be confluent, any reduction order would result in 40. The different results are an effect of leaking the substitution mapping to the outside of the lexical scope in which it was used. A simple solution could involve saving the substitution mapping once the machine enters an abstraction, then restore the same substitution mapping when it leaves. Figure 3.4 is a

$$\frac{\Theta, f \rightarrow S, (\lambda x.e) \quad S \cup \{x \mapsto z\}, e \rightarrow \Sigma, y}{\Theta, f z \rightarrow \Theta, y} \text{App}$$

Figure 3.4: An application rule which does not leak

slightly modified version of Figure 3.3b, with the difference of requiring the resulting substitution mapping to be the initial substitution mapping.

Figure 3.4 does not handle closure correctly. Evaluating Listing 3.7 under the rules defined in Figure 3.4 yields yet another problem. When line

Listing 3.7: Program with closure

```

1 fun g x =
2   fun f z = x + z; // closes x
3   fun run y =
4     let x = 20;
5     f y;
6   run 10;
7 g 3;
```

4 is reached in Listing 3.7, then  $x$  in the substitution mapping is overwritten by the rules in Figure 3.4. Even if renaming is performed such that all variables become unique, recursive functions still remain a problem. Taking a step back and considering what the evaluation strategy which involves eager substitution implies. The *first* substitution mapping which reaches an expression, is the substitution mapping which is relevant for that expression. By these semantics substitution mappings must be paired with expressions and should now be read as “ $S, e$  are the pair of the environment  $S$  necessary to evaluate the expression  $e$ ”. Substitutions bound to expressions should remain immutable; whenever an expression is discovered during interpretation, the substitution at that time is the only important substitution mapping. The  $\cup$  operator for two substitutions should now be inverted, that is,  $S \cup \Sigma$  should now pick substitutions from  $S$  if duplicates occur. Furthermore, notice that binding substitutions to expressions also allows substitutions to float up throughout interpretation. Now the

application rule must implement Equation 3.10, seen in Figure 3.5.

$$\frac{\Theta, f \rightarrow S, (\lambda x.e) \quad \{x \mapsto S, z\} \cup S, e \rightarrow \Sigma, y}{\Theta, f z \rightarrow \Sigma, y} \text{App}$$

Figure 3.5: An application rule which implements Equation 3.10 and paired substitutions. (the entire set of rules can be found in Figure 3.6)

**Remark 3.2.3.** An interesting observation is that these semantics introduce expressions with closures as a construct similar to classes in object oriented languages.  $e$  is an abstract representation of something, while the environment  $S$  is the necessary values to instantiate  $e$ , together they can produce some output. Curried functions can also be seen this way, yet it remains an interesting observation.

Surely Figure 3.5 solves the problem of lazy substitutions, but alas this is not yet the case. To reach a set of rules which will guarantee lazy substitutions and allow call by need, an additional ingredient is needed. To understand the need for this ingredient, consider the following valid state  $\{x \mapsto z, y \mapsto x\} (\lambda x.y) g$  in the context of Figure 3.5. When visiting  $(\lambda x.y)$  then the following substitution mapping is applied  $\{x \mapsto g\}$ . Once  $y$  is visited the substitution mapping will be  $\{x \mapsto g, y \mapsto x\}$ , substituting again yields  $x$  which becomes  $g$  which does not necessarily satisfy  $g \equiv z$ . One could rename all variables in some phase in compilation, but the case for recursion is not handled since a duplicate variable name can occur. Consider the program in Listing 3.8 which yields an invalid state during interpretation. Once Line 2 is reached the first time, the program will have

Listing 3.8: Recursive program with ambiguity

```

1 fun f x =
2   f (x + x);
3 f 10;
```

the state  $\{\dots, x \mapsto 10\}$ . On the second iteration, the program will have the state  $\{\dots, x \mapsto (x + x)\}$ , which does not make sense.

Taking a step back and considering what closures and partial application need to behave deterministically is rooted in solving variable ambiguity. Instead, if all variables are unique, ambiguity cannot occur. By renaming all variables to a unique name when they are introduced, the evaluation model can be simplified significantly and allows *any* of the aforementioned application rules (some of which imply eager substitution). The operation of renaming is called an  $\alpha$ -conversion.

### $\alpha$ -conversions

The  $\alpha$ -conversion mapping can occur as the substitution mapping, that is  $\{x \mapsto \gamma_1, y \mapsto \gamma_2\}$ ;  $\alpha$ -conversions are philosophical constructs more than materialized mappings.  $\alpha$ -conversions guarantee what is called  *$\alpha$ -equivalence* which is the notion of semantic equivalence. For instance  $\lambda x.x$  is  $\alpha$ -equivalent with  $\lambda \gamma.\gamma$  since both expressions are semantically equivalent.

Listing 3.9: Recursive addition function

```
1 let f = ( $\lambda f' . \lambda x .$   
2   if (x = 0) x else f' f' ((x - 1) + (x - 1))) in  
3 f f 10
```

An  $\alpha$ -conversion algorithm can be implemented such that when a new variable is introduced through an abstraction, a new name for the variable is given. More formally; Let  $V_1$  be the domain of variables in the program and  $V_2$  be the infinite domain for variable names that satisfies  $V_1 \cap V_2 = \emptyset$ , then when a new variable  $x$  is discovered, replace it with some  $\gamma \in V_2$  and let  $V_2 = V_2 \setminus \{\gamma\}$ . For instance, let  $(\lambda \gamma_1 . \lambda \gamma_2 . \text{if } (\gamma_2 = 0) \gamma_2 \text{ else } \gamma_1 \gamma_1 ((\gamma_2 - 1) + (\gamma_2 - 1)))$  be the  $\alpha$ -converted version of  $f$  in Listing 3.9.

We almost have all the building blocks in place now. The problem of ambiguity between *different* variables, that is, variables different places in the program with the same name and in recursion, can now be solved. Once again, not breaking suspension remains as an interesting value. In Figure 3.6 the **fresh** function picks a new variable from  $V_2$  and updates  $V_2$  such that picked variable no longer exists in  $V_2$ .

**Example 3.2.3.** Figure 3.7 is an example of an iteration of Listing 3.9.

Call by need is a bit more tricky since it requires more than a particular evaluation order to implement [LI88]. Substitutions can be rewritten when a variable  $x$  maps to an application such that for some substitution mapping of the form  $\{\dots, x \mapsto YE\}$  then the substituted to value is reduced. When two substitutions are combined one should always choose the most reduced of the two, which is trivial since it becomes a matter of picking the "newest" version of the substituted value. An important observation which is discussed in [LI88] is that of *object* duplication. Objects are the substituted to values which are lambda calculus terms. The term object is used to emphasize the uniqueness of lambda calculus terms. A lambda calculus expression  $E$  can be labelled with some name  $a$  when found in the evaluation process such that when some value with the label  $a$  is evaluated then all duplications of  $E$  can share the computed result, effectively adding another type of mapping. Listing 3.10 is a program which under call by name computes  $y$  two times. By assigning a unique name to the expression of  $y$ ,  $a$  for instance, the unique expression can be tracked. When  $y + y$  is

$\frac{}{S, (\lambda x.e) \rightarrow S, (\lambda x.e)} \text{ Abs}$ <p>(a)</p>	$\frac{Sx \equiv x}{S, x \rightarrow S, x} \text{ Var (terminal)}$ <p>(b)</p>
$\frac{\Theta, y \rightarrow \Sigma, e}{\{x \mapsto (\Theta, y)\} \cup S, x \rightarrow \{x \mapsto (\Theta, e)\} \cup \Sigma, e} \text{ Var}$ <p>(c)</p>	
$\frac{\Theta, f \rightarrow S, (\lambda x.e) \quad \{\gamma \mapsto (S, z), x \mapsto (\emptyset, \gamma)\} \cup S, e \rightarrow \Sigma, y \quad \gamma = \text{fresh}}{\Theta, f z \rightarrow \Sigma, y} \text{ App}$ <p>(d) An application rule which renames variables</p>	

Figure 3.6: A set of rules which support the desired semantics

computed, the first evaluation of  $y$  changes  $a$  in the substitution mapping such that  $\{\dots, a \mapsto 20\}$ . Furthermore, notice that the phrase “assigning a unique name” is exactly what  $\alpha$ -conversion does. By these rules sharing can be implemented very elegantly by also noticing that variable references are in fact, a dependency tree. Let a valid substitution mapping be  $\{x \mapsto 10 + 2, y \mapsto (x + 1), z \mapsto (y + 2)\}$  with the originating program in Equation 3.11.

$$(\lambda x. (\lambda y. (\lambda z. E) (y + 2)) (x + 1)) (10 + 20) \quad (3.11)$$

If the lambda calculus sub-program in  $E$  forces  $y$ , then the substitution mapping should be updated with  $\{y \mapsto 13\}$  which implies that  $x$  is also forced  $\{x \mapsto 12\}$ .

Observe that the App rule in Figure 3.6 is the only rule which introduces **new** variables to the substitution mapping. By using the App rule, one can introduce a variable by two different names. A transitive reference is created to the originating expression, that is, two names which point to the same reference. For some function  $f$  which duplicates expressions and some state  $\{x \mapsto (\dots, e), f \mapsto (\dots, (\lambda y. \lambda z. y+z))\}$   $f x x$  the only applicable rule is App. Using the App rule two times yields a new state  $\{x \mapsto (\dots, e), f \mapsto \dots, \gamma_1 \mapsto (\dots, x), y \mapsto (\dots, \gamma_1), \gamma_2 \mapsto (\dots, x), z \mapsto (\dots, \gamma_2)\}$ . Figure 3.8 gives clearer insight into how the expressions depend on each other. If any expression in Figure 3.8 is

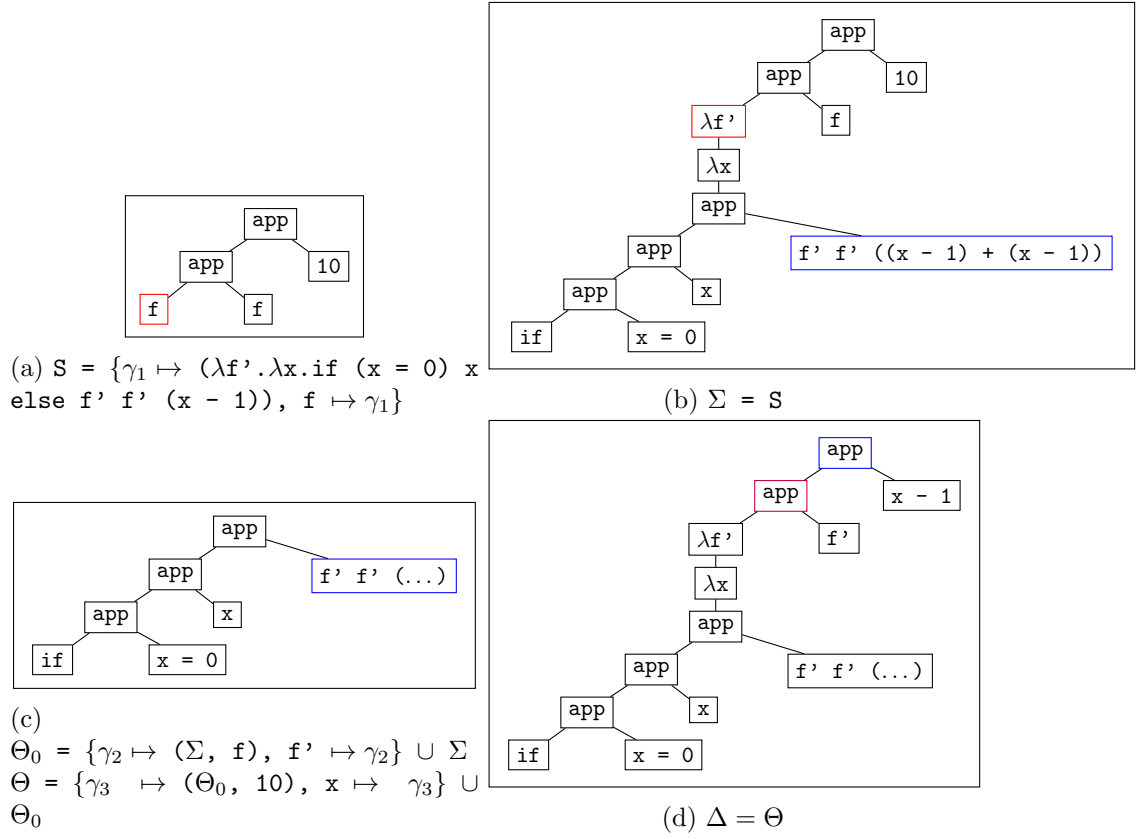


Figure 3.7: Example of substitutions in recursive function.

Listing 3.10: Program which benefits from lazy evaluation

```

1 fun double x = x + x;
2 let y = double 10;
3 fun add z k = z + k;
4 add y y;

```

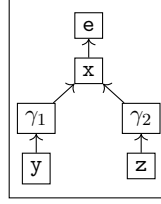


Figure 3.8: Expression dependencies

forced, then everything above will be forced. For instance if  $y$  was forced but  $z$  was not, then  $z$  would have to evaluate the reference to  $\gamma_2$  which points to  $x$  which has been evaluated and added to the substitution mapping by Figure 3.6c since  $x$  is a (transitive) parent of  $y$ . We have now arrived at semantics which are very similar to the semantics described in [Lau93] (except concepts such as lazy substitutions).

**Remark 3.2.4.** Sharing can also be implemented in a way which promotes more imperative constructs, such as variables. When some object  $E$  is discovered it is declared as a mutable pointer by the interpreter (or natively if implemented in a lazy language) such that all duplications effectively point to the same object. An implementation which builds upon pointers are explored in Chapter 12 in [Pey87].

### Garbage collection

Garbage collection in purely functional languages is different from imperative languages like C, since manual memory management is often abstracted away. Adding mappings from variables to lambda terms without any subtractions will lead to an ever increasing substitution mapping size. When performing reductions one might desire to garbage collect the substitution mapping once it contains unreachable substitutions. Fortunately the task of collecting garbage is quite easy solvable in by a naive algorithm, since the “scope” of a substitution is only relevant where the substituted variables has been introduced, which may only happen in the abstraction term. Unfortunately not all programs “return”, like programs which implement infinite recursion.

**Remark 3.2.5.** Continually passing style programs clearly need some type of garbage collection. Generally for tail call optimization (Remark 3.2.7) to be feasible, some sort of ad-hoc freeing of substitutions must be present, such that infinite programs can run in finite space.

[Lau93] presents an elegant (and abstract) rule for garbage collection (Figure 3.9, and a version with lazy substitutions Figure 3.10). The function  $R$  scans the substitution mapping  $\Theta$  for all variables in the expression  $e$ .

Problems in the domain of computer science (especially algorithms), can

$$\frac{\Theta, e \rightarrow S, y \quad x \notin R(\Theta, e)}{\Theta \cup \{x \mapsto z\}, e \rightarrow S, y} \text{GC}$$

Figure 3.9: Launchbury garbage collection

$$\frac{\Theta, e \rightarrow S, y \quad x \notin R(\Theta, e)}{\Theta \cup \{x \mapsto (\Delta, z)\}, e \rightarrow S, y} \text{GC}$$

Figure 3.10: Launchbury garbage collection with lazy substitution

usually be solved much more efficiently if the problem has certain properties. Launchbury garbage collection argues for the possibility of parallel garbage collection, which is a property granted for free in immutable languages. In imperative languages garbage collection algorithms either block the execution of the program or have complicated implementation details. Fortunately in functional programming languages, data is (traditionally) immutable, which implies that references *never* change.

The lazy substitution semantics might appear foreign to the rules of garbage collection but can indeed be beneficial in regard to the function  $R$ . The implementation of the function  $R$  is very open for interpretation. A garbage collection algorithm which collects every “lost reference” which adheres to  $R$  in Figure 3.9, could be implemented by the use of **Free** or **Bound**. A garbage collection algorithm for Figure 3.10 must also inspect the bound substitution mappings as in Equation 3.12.

$$\begin{aligned} R(\Theta, (\lambda x. e)) &= R(\Theta \setminus \{x\}, e) & (3.12) \\ R(\Theta, (f \ e)) &= R(\Theta, f) \cup R(\Theta, e) \\ R(\Theta, x) &= \{x \mapsto (e, R(S, e))\} \cup R(S, e) \quad , x \mapsto (e, S) \in \Theta \\ R(\Theta, x) &= \{\} \quad , x \mapsto (e, S) \notin \Theta \end{aligned}$$

Garbage collection algorithms often run in worst case  $O(n)$  and amortized  $O(1)$ . The algorithm proposed in Equation 3.12 also runs in similar time. This is quite simple to prove.

### 3.2.4 An invariant on infinite programs

An important problem still remains which is that of infinite programs. Imperative programming languages often solve this by introducing loops, whereas functional programming languages use recursion. Recursion may be equally powerful in terms of expressiveness, but becomes a bit more

tricky when considering interpreter details. A prerequisite for an infinitely running program to exist in practice is that the program must not grow its resource needs as it runs.

The distinction between recursive functions and loops in imperative programming languages is often what makes infinite programs expressible. In a traditional imperative language, a function allocates a *stack frame* and is explicitly parameterized, whereas a loop acts more like an anonymous closure which is always parameterized with itself (a function which is wrapped in a fixed point combinator, like the Y-combinator).

**Remark 3.2.6.** A call stack is a stack of stack frames. A stack frame is a pointer to a function pointer. Stack frames are used to return execution to the previous function (the calling function). Every time a new function is called, the called-from function places a “resume execution from here” pointer onto the call stack.

Imperative languages are also often evaluated under call by value which further simplifies implementation details. Imperative loops (more interestingly, infinite loops) can safely release all static resources (variables bindings), which were allocated in the iteration, once an iteration has completed. In traditional imperative languages recursive functions can only iterate a finite number of times, more specifically until the call stack is full.

To really understand what happens in a lambda calculus interpreter, we must understand what happens in Listing 3.11. Listing 3.11 implements two variants of a `fold` function which accumulates a list of type `List a` to a `b`. The two variants differ when considering evaluation strategy and *tail call optimization*.

**Remark 3.2.7.** Tail call optimization is an optimization which can be performed on programs with a particular structure. If the last expression is a function invocation, then the rewritten program does not grow. For instance the expression `let f = (λg.λx.g x) in ... f g' 0` is eventually rewritten to `g' 0`. If for instance the expression awaited a result like in `let f = (λg.λx.x + (g x)) in ... f g' 0`, then it would be rewritten to `x + (g' 0)`, increases the size of the program by `x +`, since the `+` operator requires both expressions to be evaluated. It should become clear that reduction strategies always imply tail call optimization, whenever possible.

The first flavor of `fold`; `foldl`, implements `fold` such that the program expression tree does not grow throughout program interpretation, under a call by value environment. The constraint on evaluation strategy is important for `foldl`, for reasons which will become clear once other evaluation



Listing 3.11: Program that implements two functions that fold a List `a` to `a b`

```
1 type List a =  
2   | Nil  
3   | Cons a (List a)  
4 ;  
5 fun add a b = a + b;  
6  
7 fun foldl f z l =  
8   match l  
9     | Nil -> z;  
10    | Cons x xs ->  
11      foldl f (f x z) xs;  
12 ;  
13  
14 fun foldr f z l =  
15   match l  
16     | Nil -> z;  
17     | Cons x xs ->  
18       f x (foldr f z xs);  
19 ;
```

strategies are discussed.

$$\begin{aligned}
& \text{foldl } 0 \text{ add } (\text{Cons } 1 (\text{Cons } 2 \dots (\text{Cons } n \text{ Nil}))) & (3.13) \\
= & 1 \text{ z } (\lambda \text{xs}, \text{x}. \text{foldl } f \text{ (f x z) xs}) \\
= & (\text{Cons } 1 (\text{Cons } 2 \dots (\text{Cons } n \text{ Nil}))) \text{ z } (\lambda \text{xs}, \text{x}. \text{foldl } f \text{ (f x z) xs}) \\
= & \text{foldl } f \text{ (f x z) xs } \{ \text{xs} \mapsto (\text{Cons } 2 (\text{Cons } 3 \dots (\text{Cons } n \text{ Nil}))), \text{x} \mapsto 1, \dots \} \\
= & \text{foldl } \text{add} (\text{add } 1 \text{ } 0) (\text{Cons } 2 (\text{Cons } 3 \dots (\text{Cons } n \text{ Nil}))) \{ \dots \} \\
= & \text{foldl } \text{add } 1 (\text{Cons } 2 (\text{Cons } 3 \dots (\text{Cons } n \text{ Nil}))) \{ \dots \} \\
& \dots
\end{aligned}$$

Evaluating `foldl` on a list of size `n` with the addition function showcases how the program only grows by a constant number of terms.

**Remark 3.2.8.** Note again that the list is always refereed to by reference; the list is not copied.

## Part II

# Algorithms and Datastructures

## Chapter 4

# Conventional data structures and terminology

Data structures in traditional contexts are *homogeneous* collections of data, usually with a particular shape represented by an algebraic data type, with an associated set of *morphisms*. A homogeneous collection of data is a collection in which every element is of the same type. Morphisms come in various forms, they essentially encapsulate the operations that can be performed on a data structure (or more generally an object). Algebraic data structures and their associated morphisms come together into an algebra.

**Remark 4.0.1.** In object oriented programming data structures (an algebra) is most often implemented through a class while functional programming languages often separate the shape and operations.

Conventional data structures encapsulates data structures which are interesting under the call by value (section 3.1) evaluation strategy. Evaluation strategies have many implications on the data structure in question. In call by name or call by need one would have to be careful not to create an unnecessary dependency which may force a computation which could otherwise stay suspended. The choice of evaluation strategy and data structure implementation has a significant impact on complexity analysis, which will be explored.

### 4.1 Lists and lazy evaluation

An instance of a data structure which has been thoroughly discussed throughout this thesis is `List` (Listing 1.17). `List` is an excellent choice as an introductory data structure since it gives insight into some very universal problems regarding both immutable and mutable data structures. One is free to choose the operations for `List` but a common operation is `map` (Listing 4.1).

Listing 4.1: Mapping from List *a* to List *b*

```

1 fun map f l =
2   match l
3   | Cons x xs -> Cons (f x) (map f xs);
4   | Nil -> l;
5   ;

```

There exists several analytical techniques to justify performance guarantees in call by value data structures, the most straight forward of which is the worst case analysis. Worst case analysis is usually the most straight forward, since it becomes a matter of finding the worst input for any possible state of the data structure.

An interesting observation from `map` is that it runs differently in a call by need environment compared to a call by value environment. In a call by value environment `map` takes  $\Theta(n)$  time since every `Cons`'ed value must be visited. In a call by need environment things become a bit more philosophical. When `map` is evaluated in a call by need environment it is technically suspended thus always requires one operation. When a value which depends on *one* `map` invocation, is forced (from the addition operator for instance), then the computational complexity has the same bounds as if it were call by value. The computational complexity in a call by need (or name) environment for *one* `map` invocation is thus  $O(n)$  and  $\Omega(1)$ , in general *all* call by need algorithms run in  $\Omega(1)$ . More interestingly, consider  $n$  invocations of `map` (Listing 4.2) on some list.

Listing 4.2:  $n$  invocation of `map`

```

1 fun id x = x;
2 let xs = Cons 1 (Cons 2 (... (Cons m Nil)));
3 let m1 = map id xs;
4 let m2 = map id m1;
5 ...
6 let mn = map id mn-1;

```

Clearly  $m_n$  in Listing 4.2 requires  $n \cdot m$  time if it is forced. Moreover observe that we can “enqueue” an unbounded amount of `map` operations, or rather, the computational complexity is not a function of  $n$  (a function of the input size), but rather a function of how much work has been performed on the data structure.

With bounds such as  $\Omega(1)$  and a worst case which is unbounded, traditional worst case analysis breaks down.

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## Chapter 5

# Appendix

Listing 5.1: The output of an exponential type

```
1 ##### tuple #####
2 substitution set Map(c0 -> (a0 -> (b0 -> d0)))
3 type (a0 -> (b0 -> ((a0 -> (b0 -> d0)) -> d0)))
4 Type vars in sub = 4
5 ##### tuple #####
6 ##### one #####
7 substitution set Map(c0 -> (a0 -> (b0 -> d0)), e0 ->
  (h0 -> (i0 -> ((h0 -> (i0 -> j0)) -> j0))), f0
  -> (k0 -> (l0 -> ((k0 -> (l0 -> m0)) -> m0))), n0
  -> (((h0 -> (i0 -> ((h0 -> (i0 -> j0)) -> j0)))
  -> ((k0 -> (l0 -> ((k0 -> (l0 -> m0)) -> m0))) ->
  g0)) ->
8 g0))
9 type (((h0 -> (i0 -> ((h0 -> (i0 -> j0)) -> j0))) ->
  ((k0 -> (l0 -> ((k0 -> (l0 -> m0)) -> m0))) ->
  g0)) -> g0)
10 current env is Map(tuple -> Scheme(Set(a0, b0, d0), (
  a0 -> (b0 -> ((a0 -> (b0 -> d0)) -> d0))), one
  -> Scheme(Set(k0, g0, h0, i0, l0, m0, j0), (((h0
  -> (i0 -> ((h0 -> (i0 -> j0)) -> j0))) -> ((k0 ->
  (l0 -> ((k0 -> (l0 -> m0)) -> m0))) -> g0)) ->
  g0)))
11 Type vars = 14
12 Type vars in sub = 33
13 ##### one #####
14 ##### two #####
15 substitution set Map(o0 -> (((t0 -> (u0 -> ((t0 -> (
  u0 -> x0)) -> x0))) -> ((r0 -> (v0 -> ((r0 -> (v0
  -> w0)) -> w0))) -> s0)) -> s0), e0 -> (h0 -> (
```



```

16   i0 -> ((h0 -> (i0 -> j0)) -> j0))), f0 -> (k0 ->
      (l0 -> ((k0 -> (l0 -> m0)) -> m0))), p0 -> (((a1
      -> (b1 ->
17   ((a1 -> (b1 -> e1)) -> e1))) -> ((y0 -> (c1 -> ((y0
      -> (c1 -> d1)) -> d1))) -> z0)) -> z0), n0 ->
      (((h0 -> (i0 -> ((h0 -> (i0 -> j0)) -> j0))) ->
      ((k0 -> (l0 -> ((k0 -> (l0 -> m0)) -> m0))) ->
      g0)) -> g0), c0 -> (a0 -> (b0 -> d0)), f1 ->
      (((((t0 -> (u0 ->
18   ((t0 -> (u0 -> x0)) -> x0))) -> ((r0 -> (v0 -> ((r0
      -> (v0 -> w0)) -> w0))) -> s0)) -> s0) -> (((
      a1 -> (b1 -> ((a1 -> (b1 -> e1)) -> e1))) -> ((
      y0 -> (c1 -> ((y0 -> (c1 -> d1)) -> d1))) -> z0)
      ) -> z0) -> q0)) -> q0))
19 type (((((t0 -> (u0 -> ((t0 -> (u0 -> x0)) -> x0)))
      -> ((r0 -> (v0 -> ((r0 -> (v0 -> w0)) -> w0))) ->
      s0)) -> s0) -> (((a1 -> (b1 -> ((a1 -> (b1 ->
      e1)) -> e1))) -> ((y0 -> (c1 -> ((y0 -> (c1 -> d1)
      )) -> d1))) -> z0)) -> z0) -> q0)) -> q0)
20 current env is Map(tuple -> Scheme(Set(a0, b0, d0),(
      a0 -> (b0 -> ((a0 -> (b0 -> d0)) -> d0))), one
      -> Scheme(Set(k0, g0, h0, i0, l0, m0, j0),(((h0
      -> (i0 -> ((h0 -> (i0 -> j0)) -> j0))) -> ((k0 ->
      (l0 -> ((k0 -> (l0 -> m0)) -> m0))) -> g0)) ->
      g0)), two -
21 > Scheme(Set(u0, x0, q0, a1, b1, s0, e1, d1, z0, w0,
      y0, v0, t0, c1, r0),((((t0 -> (u0 -> ((t0 -> (
      u0 -> x0)) -> x0))) -> ((r0 -> (v0 -> ((r0 -> (v0
      -> w0)) -> w0))) -> s0)) -> s0) -> (((a1 -> (b1
      -> ((a1 -> (b1 -> e1)) -> e1))) -> ((y0 -> (c1
      -> ((y0 ->
22   (c1 -> d1)) -> d1))) -> z0)) -> z0) -> q0)) -> q0))
      )
23 Type vars = 32
24 Type vars in sub = 94
25 ##### two #####
26 substitution set Map(o0 -> (((t0 -> (u0 -> ((t0 -> (
      u0 -> x0)) -> x0))) -> ((r0 -> (v0 -> ((r0 -> (v0
      -> w0)) -> w0))) -> s0)) -> s0), e0 -> (h0 -> (
      i0 -> ((h0 -> (i0 -> j0)) -> j0))), n2 -> (((((((
      v1 -> (j1 -> ((v1 -> (j1 -> k1)) -> k1))) -> ((x1
      -> (u1 -

```

```

27 > ((x1 -> (u1 -> s1)) -> s1))) -> o1)) -> o1) ->
    (((m1 -> (n1 -> ((m1 -> (n1 -> p1)) -> p1))) ->
    ((t1 -> (w1 -> ((t1 -> (w1 -> q1)) -> q1))) -> r1
    )) -> r1) -> l1)) -> l1) -> ((((((k2 -> (y1 -> ((
    k2 -> (y1 -> z1)) -> z1))) -> ((m2 -> (j2 -> ((m2
    -> (j2 ->
28 h2)) -> h2))) -> d2)) -> d2) -> (((b2 -> (c2 -> ((
    b2 -> (c2 -> e2)) -> e2))) -> ((i2 -> (l2 -> ((i2
    -> (l2 -> f2)) -> f2))) -> g2)) -> g2) -> a2))
    -> a2) -> i1)) -> i1), f0 -> (k0 -> (l0 -> ((k0
    -> (l0 -> m0)) -> m0))), p0 -> (((a1 -> (b1 -> ((
    a1 -> (b1 -
29 > e1)) -> e1))) -> ((y0 -> (c1 -> ((y0 -> (c1 -> d1)
    ) -> d1))) -> z0)) -> z0), n0 -> (((h0 -> (i0 ->
    ((h0 -> (i0 -> j0)) -> j0))) -> ((k0 -> (l0 -> ((
    k0 -> (l0 -> m0)) -> m0))) -> g0)) -> g0), c0 ->
    (a0 -> (b0 -> d0)), g1 -> (((((v1 -> (j1 -> ((v1
    -> (j1 -
30 > k1)) -> k1))) -> ((x1 -> (u1 -> ((x1 -> (u1 -> s1)
    ) -> s1))) -> o1)) -> o1) -> (((m1 -> (n1 -> ((
    m1 -> (n1 -> p1)) -> p1))) -> ((t1 -> (w1 -> ((t1
    -> (w1 -> q1)) -> q1))) -> r1)) -> r1) -> l1))
    -> l1), h1 -> (((((k2 -> (y1 -> ((k2 -> (y1 -> z1
    )) -> z1))
31 ) -> ((m2 -> (j2 -> ((m2 -> (j2 -> h2)) -> h2))) ->
    d2)) -> d2) -> (((b2 -> (c2 -> ((b2 -> (c2 -> e2
    )) -> e2))) -> ((i2 -> (l2 -> ((i2 -> (l2 -> f2))
    -> f2))) -> g2)) -> g2) -> a2)) -> a2), f1 ->
    (((((t0 -> (u0 -> ((t0 -> (u0 -> x0)) -> x0))) ->
    ((r0 -> (
32 v0 -> ((r0 -> (v0 -> w0)) -> w0))) -> s0)) -> s0) ->
    (((a1 -> (b1 -> ((a1 -> (b1 -> e1)) -> e1))) ->
    ((y0 -> (c1 -> ((y0 -> (c1 -> d1)) -> d1))) ->
    z0)) -> z0) -> q0)) -> q0))
33 type (((((((v1 -> (j1 -> ((v1 -> (j1 -> k1)) -> k1))
    ) -> ((x1 -> (u1 -> ((x1 -> (u1 -> s1)) -> s1)))
    -> o1)) -> o1) -> (((m1 -> (n1 -> ((m1 -> (n1 ->
    p1)) -> p1))) -> ((t1 -> (w1 -> ((t1 -> (w1 ->
    q1)) -> q1))) -> r1)) -> r1) -> l1)) -> l1) ->
    ((((((k2 ->
34 (y1 -> ((k2 -> (y1 -> z1)) -> z1))) -> ((m2 -> (j2
    -> ((m2 -> (j2 -> h2)) -> h2))) -> d2)) -> d2)
    -> (((b2 -> (c2 -> ((b2 -> (c2 -> e2)) -> e2)))
    -> ((i2 -> (l2 -> ((i2 -> (l2 -> f2)) -> f2)))

```

```

    -> g2)) -> g2) -> a2)) -> a2) -> i1)) -> i1)
35 current env is Map(tuple -> Scheme(Set(a0, b0, d0), (
    a0 -> (b0 -> ((a0 -> (b0 -> d0)) -> d0))), one
    -> Scheme(Set(k0, g0, h0, i0, l0, m0, j0), (((h0
    -> (i0 -> ((h0 -> (i0 -> j0)) -> j0))) -> ((k0 ->
    (l0 -> ((k0 -> (l0 -> m0)) -> m0))) -> g0)) ->
    g0)), two -
36 > Scheme(Set(u0, x0, q0, a1, b1, s0, e1, d1, z0, w0,
    y0, v0, t0, c1, r0), (((((t0 -> (u0 -> ((t0 -> (
    u0 -> x0)) -> x0))) -> ((r0 -> (v0 -> ((r0 -> (v0
    -> w0)) -> w0))) -> s0)) -> s0) -> (((a1 -> (b1
    -> ((a1 -> (b1 -> e1)) -> e1))) -> ((y0 -> (c1
    -> ((y0 ->
37 (c1 -> d1)) -> d1))) -> z0)) -> z0) -> q0)) -> q0))
    , three -> Scheme(Set(j2, i1, m1, c2, n1, r1, z1
    , g2, q1, s1, w1, t1, f2, x1, o1, j1, i2, a2, h2
    , b2, u1, v1, p1, k2, m2, l2, l1, y1, e2, k1, d2
    ), (((((((v1 -> (j1 -> ((v1 -> (j1 -> k1)) -> k1)
    )) -> ((x1 -
38 > (u1 -> ((x1 -> (u1 -> s1)) -> s1))) -> o1)) -> o1)
    -> (((m1 -> (n1 -> ((m1 -> (n1 -> p1)) -> p1)))
    -> ((t1 -> (w1 -> ((t1 -> (w1 -> q1)) -> q1)))
    -> r1)) -> r1) -> l1)) -> l1) -> ((((((k2 -> (y1
    -> ((k2 -> (y1 -> z1)) -> z1))) -> ((m2 -> (j2 ->
    ((m2 ->
39 (j2 -> h2)) -> h2))) -> d2)) -> d2) -> (((b2 -> (c2
    -> ((b2 -> (c2 -> e2)) -> e2))) -> ((i2 -> (l2
    -> ((i2 -> (l2 -> f2)) -> f2))) -> g2)) -> g2) ->
    a2)) -> a2) -> i1)) -> i1)))
40 Type vars = 66
41 Type vars in sub = 219
42 ##### three #####
43 ##### four #####
44 substitution set Map(o0 -> (((t0 -> (u0 -> ((t0 -> (
    u0 -> x0)) -> x0))) -> ((r0 -> (v0 -> ((r0 -> (v0
    -> w0)) -> w0))) -> s0)) -> s0), b5 -> ((((((((((
    m3 -> (g3 -> ((m3 -> (g3 -> u3)) -> u3))) -> ((e3
    -> (l3 -> ((e3 -> (l3 -> a3)) -> a3))) -> f3))
    -> f3) ->
45 (((((t2 -> (v2 -> ((t2 -> (v2 -> n3)) -> n3))) -> ((
    c3 -> (b3 -> ((c3 -> (b3 -> z2)) -> z2))) -> w2)
    ) -> w2) -> r3)) -> r3) -> (((((((o3 -> (s3 -> ((
    o3 -> (s3 -> x2)) -> x2))) -> ((p3 -> (r2 -> ((
    p3 -> (r2 -> j3)) -> j3))) -> v3)) -> v3) ->

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46      (((k3 -> (u2 ->
      ((k3 -> (u2 -> t3)) -> t3))) -> ((h3 -> (q3 -> ((h3
        -> (q3 -> d3)) -> d3))) -> y2)) -> y2) -> i3))
        -> i3) -> s2)) -> s2) -> (((((((r4 -> (l4 -> ((
          r4 -> (l4 -> z4)) -> z4))) -> ((j4 -> (q4 -> ((
            j4 -> (q4 -> f4)) -> f4))) -> k4)) -> k4) ->
          (((y3 -> (a4 ->
47      ((y3 -> (a4 -> s4)) -> s4))) -> ((h4 -> (g4 -> ((h4
        -> (g4 -> e4)) -> e4))) -> b4)) -> b4) -> w4))
          -> w4) -> ((((((t4 -> (x4 -> ((t4 -> (x4 -> c4))
            -> c4))) -> ((u4 -> (w3 -> ((u4 -> (w3 -> o4))
              -> o4))) -> a5)) -> a5) -> (((p4 -> (z3 -> ((p4
                -> (z3 -> y
48      4)) -> y4))) -> ((m4 -> (v4 -> ((m4 -> (v4 -> i4))
          -> i4))) -> d4)) -> d4) -> n4)) -> n4) -> x3)) ->
            x3) -> q2)) -> q2), o2 -> (((((((m3 -> (g3 -> ((
              m3 -> (g3 -> u3)) -> u3))) -> ((e3 -> (l3 -> ((e3
                -> (l3 -> a3)) -> a3))) -> f3)) -> f3) -> (((t2
                  -> (v2 -
49      > ((t2 -> (v2 -> n3)) -> n3))) -> ((c3 -> (b3 -> ((
          c3 -> (b3 -> z2)) -> z2))) -> w2)) -> w2) -> r3))
            -> r3) -> ((((((o3 -> (s3 -> ((o3 -> (s3 -> x2))
              -> x2))) -> ((p3 -> (r2 -> ((p3 -> (r2 -> j3))
                -> j3))) -> v3)) -> v3) -> (((k3 -> (u2 -> ((k3
                  -> (u2 ->
50      t3)) -> t3))) -> ((h3 -> (q3 -> ((h3 -> (q3 -> d3))
          -> d3))) -> y2)) -> y2) -> i3)) -> i3) -> s2)) ->
            s2), e0 -> (h0 -> (i0 -> ((h0 -> (i0 -> j0)) ->
              j0))), n2 -> (((((((v1 -> (j1 -> ((v1 -> (j1 ->
                k1)) -> k1))) -> ((x1 -> (u1 -> ((x1 -> (u1 -> s1
                  )) -> s1)
51      )) -> o1)) -> o1) -> (((m1 -> (n1 -> ((m1 -> (n1 ->
          p1)) -> p1))) -> ((t1 -> (w1 -> ((t1 -> (w1 ->
            q1)) -> q1))) -> r1)) -> r1) -> l1)) -> l1) ->
            (((((((k2 -> (y1 -> ((k2 -> (y1 -> z1)) -> z1)))
              -> ((m2 -> (j2 -> ((m2 -> (j2 -> h2)) -> h2))) ->
                d2)) -> d2)
52      -> (((b2 -> (c2 -> ((b2 -> (c2 -> e2)) -> e2))) ->
          ((i2 -> (l2 -> ((i2 -> (l2 -> f2)) -> f2))) ->
            g2)) -> g2) -> a2)) -> a2) -> i1)) -> i1), f0 ->
            (k0 -> (l0 -> ((k0 -> (l0 -> m0)) -> m0))), p0
            -> (((a1 -> (b1 -> ((a1 -> (b1 -> e1)) -> e1)))
              -> ((y0 -> (

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53 c1 -> ((y0 -> (c1 -> d1)) -> d1))) -> z0)) -> z0),
    n0 -> (((h0 -> (i0 -> ((h0 -> (i0 -> j0)) -> j0))
    ) -> ((k0 -> (l0 -> ((k0 -> (l0 -> m0)) -> m0)))
    -> g0)) -> g0), c0 -> (a0 -> (b0 -> d0)), g1 ->
    (((((v1 -> (j1 -> ((v1 -> (j1 -> k1)) -> k1))) ->
    ((x1 -> (
54 u1 -> ((x1 -> (u1 -> s1)) -> s1))) -> o1)) -> o1) ->
    (((m1 -> (n1 -> ((m1 -> (n1 -> p1)) -> p1))) ->
    ((t1 -> (w1 -> ((t1 -> (w1 -> q1)) -> q1))) ->
    r1)) -> r1) -> l1)) -> l1), h1 -> (((((k2 -> (y1
    -> ((k2 -> (y1 -> z1)) -> z1))) -> ((m2 -> (j2 ->
    ((m2 ->
55 (j2 -> h2)) -> h2))) -> d2)) -> d2) -> (((b2 -> (c2
    -> ((b2 -> (c2 -> e2)) -> e2))) -> ((i2 -> (l2
    -> ((i2 -> (l2 -> f2)) -> f2))) -> g2)) -> g2) ->
    a2)) -> a2), p2 -> (((((((r4 -> (l4 -> ((r4 -> (
    l4 -> z4)) -> z4))) -> ((j4 -> (q4 -> ((j4 -> (q4
    -> f4))
56 -> f4))) -> k4)) -> k4) -> (((y3 -> (a4 -> ((y3 ->
    (a4 -> s4)) -> s4))) -> ((h4 -> (g4 -> ((h4 -> (
    g4 -> e4)) -> e4))) -> b4)) -> b4) -> w4)) -> w4)
    -> (((((((t4 -> (x4 -> ((t4 -> (x4 -> c4)) -> c4)
    )) -> ((u4 -> (w3 -> ((u4 -> (w3 -> o4)) -> o4)))
    -> a5))
57 -> a5) -> (((p4 -> (z3 -> ((p4 -> (z3 -> y4)) -> y4
    ))) -> ((m4 -> (v4 -> ((m4 -> (v4 -> i4)) -> i4))
    ) -> d4)) -> d4) -> n4)) -> n4) -> x3)) -> x3),
    f1 -> (((((t0 -> (u0 -> ((t0 -> (u0 -> x0)) -> x0
    ))) -> ((r0 -> (v0 -> ((r0 -> (v0 -> w0)) -> w0))
    ) -> s0))
58 -> s0) -> (((a1 -> (b1 -> ((a1 -> (b1 -> e1)) ->
    e1))) -> ((y0 -> (c1 -> ((y0 -> (c1 -> d1)) ->
    d1))) -> z0)) -> z0) -> q0)) -> q0))
59 type (((((((m3 -> (g3 -> ((m3 -> (g3 -> u3)) -> u3
    ))) -> ((e3 -> (l3 -> ((e3 -> (l3 -> a3)) -> a3))
    ) -> f3)) -> f3) -> (((t2 -> (v2 -> ((t2 -> (v2
    -> n3)) -> n3))) -> ((c3 -> (b3 -> ((c3 -> (b3 ->
    z2)) -> z2))) -> w2)) -> w2) -> r3)) -> r3) ->
    ((((((o3
60 -> (s3 -> ((o3 -> (s3 -> x2)) -> x2))) -> ((p3 -> (
    r2 -> ((p3 -> (r2 -> j3)) -> j3))) -> v3)) -> v3)
    -> (((k3 -> (u2 -> ((k3 -> (u2 -> t3)) -> t3)))
    -> ((h3 -> (q3 -> ((h3 -> (q3 -> d3)) -> d3)))
    -> y2)) -> y2) -> i3)) -> i3) -> s2)) -> s2) ->

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61      (((((((r4
-> (l4 -> ((r4 -> (l4 -> z4)) -> z4))) -> ((j4 -> (
q4 -> ((j4 -> (q4 -> f4)) -> f4))) -> k4)) -> k4)
-> (((((y3 -> (a4 -> ((y3 -> (a4 -> s4)) -> s4)))
-> ((h4 -> (g4 -> ((h4 -> (g4 -> e4)) -> e4)))
-> b4)) -> b4) -> w4)) -> w4) -> (((((((t4 -> (x4
-> ((t4 ->
62 (x4 -> c4)) -> c4))) -> ((u4 -> (w3 -> ((u4 -> (w3
-> o4)) -> o4))) -> a5)) -> a5) -> (((p4 -> (z3
-> ((p4 -> (z3 -> y4)) -> y4))) -> ((m4 -> (v4
-> ((m4 -> (v4 -> i4)) -> i4))) -> d4)) -> d4)
-> n4)) -> n4) -> x3)) -> x3) -> q2)) -> q2)
63 current env is Map(four -> Scheme(Set(s4, q4, y4, d3
, t4, g3, r2, k3, w2, x3, y2, r3, c3, m4, i4, w3,
v4, u4, u3, w4, r4, z2, i3, u2, y3, a5, s2, g4,
f3, t2, n3, l3, v3, c4, f4, x4, x2, h4, j3, t3,
z4, e3, m3, n4, h3, v2, s3, e4, o4, z3, k4, b4,
b3, j4, a3
64 , q2, l4, q3, a4, p4, p3, d4, o3),((((((((m3 -> (g3
-> ((m3 -> (g3 -> u3)) -> u3))) -> ((e3 -> (l3
-> ((e3 -> (l3 -> a3)) -> a3))) -> f3)) -> f3) ->
((((t2 -> (v2 -> ((t2 -> (v2 -> n3)) -> n3))) ->
((c3 -> (b3 -> ((c3 -> (b3 -> z2)) -> z2))) ->
w2)) -> w2
65 ) -> r3)) -> r3) -> (((((((o3 -> (s3 -> ((o3 -> (s3
-> x2)) -> x2))) -> ((p3 -> (r2 -> ((p3 -> (r2 ->
j3)) -> j3))) -> v3)) -> v3) -> (((k3 -> (u2 ->
((k3 -> (u2 -> t3)) -> t3))) -> ((h3 -> (q3 ->
((h3 -> (q3 -> d3)) -> d3))) -> y2)) -> y2) -> i3
)) -> i3)
66 -> s2)) -> s2) -> (((((((r4 -> (l4 -> ((r4 -> (l4
-> z4)) -> z4))) -> ((j4 -> (q4 -> ((j4 -> (q4 ->
f4)) -> f4))) -> k4)) -> k4) -> (((((y3 -> (a4 ->
((y3 -> (a4 -> s4)) -> s4))) -> ((h4 -> (g4 ->
((h4 -> (g4 -> e4)) -> e4))) -> b4)) -> b4) -> w4
)) -> w4)
67 -> (((((((t4 -> (x4 -> ((t4 -> (x4 -> c4)) -> c4)))
-> ((u4 -> (w3 -> ((u4 -> (w3 -> o4)) -> o4))) ->
a5)) -> a5) -> (((p4 -> (z3 -> ((p4 -> (z3 ->
y4)) -> y4))) -> ((m4 -> (v4 -> ((m4 -> (v4 -> i4
)) -> i4))) -> d4)) -> d4) -> n4)) -> n4) -> x3))
-> x3) ->
68 q2)) -> q2)), three -> Scheme(Set(j2, i1, m1, c2,
n1, r1, z1, g2, q1, s1, w1, t1, f2, x1, o1, j1,

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        i2, a2, h2, b2, u1, v1, p1, k2, m2, l2, l1, y1,
        e2, k1, d2),(((((((v1 -> (j1 -> ((v1 -> (j1 ->
        k1)) -> k1))) -> ((x1 -> (u1 -> ((x1 -> (u1 ->
        s1)) -> s1)))) -
69 > o1)) -> o1) -> (((((m1 -> (n1 -> ((m1 -> (n1 -> p1)
        ) -> p1))) -> ((t1 -> (w1 -> ((t1 -> (w1 -> q1))
        -> q1))) -> r1)) -> r1) -> l1)) -> l1) -> ((((((
        k2 -> (y1 -> ((k2 -> (y1 -> z1)) -> z1))) -> ((m2
        -> (j2 -> ((m2 -> (j2 -> h2)) -> h2))) -> d2))
        -> d2) ->
70 (((((b2 -> (c2 -> ((b2 -> (c2 -> e2)) -> e2))) -> ((
        i2 -> (l2 -> ((i2 -> (l2 -> f2)) -> f2))) -> g2))
        -> g2) -> a2)) -> a2) -> i1)) -> i1)), two ->
        Scheme(Set(u0, x0, q0, a1, b1, s0, e1, d1, z0, w0
        , y0, v0, t0, c1, r0),((((((t0 -> (u0 -> ((t0 -> (
        u0 -> x0))
71 -> x0))) -> ((r0 -> (v0 -> ((r0 -> (v0 -> w0)) -> w0
        ))) -> s0)) -> s0) -> (((((a1 -> (b1 -> ((a1 -> (
        b1 -> e1)) -> e1))) -> ((y0 -> (c1 -> ((y0 -> (c1
        -> d1)) -> d1))) -> z0)) -> z0) -> q0)) -> q0)),
        tuple -> Scheme(Set(a0, b0, d0),(a0 -> (b0 -> ((
        a0 -> (b0
72 -> d0)) -> d0))))), one -> Scheme(Set(k0, g0, h0, i0
        , l0, m0, j0),((((h0 -> (i0 -> ((h0 -> (i0 -> j0)
        ) -> j0))) -> ((k0 -> (l0 -> ((k0 -> (l0 -> m0))
        -> m0))) -> g0)) -> g0)))
73 Type vars = 132
74 Type vars in sub = 472
75 ##### four #####
76 ##### main #####
77 substitution set Map(o0 -> (((t0 -> (u0 -> ((t0 -> (
        u0 -> x0)) -> x0))) -> ((r0 -> (v0 -> ((r0 -> (v0
        -> w0)) -> w0))) -> s0)) -> s0), b5 -> ((((((((((
        m3 -> (g3 -> ((m3 -> (g3 -> u3)) -> u3))) -> ((e3
        -> (l3 -> ((e3 -> (l3 -> a3)) -> a3))) -> f3))
        -> f3) ->
78 (((((t2 -> (v2 -> ((t2 -> (v2 -> n3)) -> n3))) -> ((
        c3 -> (b3 -> ((c3 -> (b3 -> z2)) -> z2))) -> w2)
        ) -> w2) -> r3)) -> r3) -> (((((((o3 -> (s3 -> ((
        o3 -> (s3 -> x2)) -> x2))) -> ((p3 -> (r2 -> ((
        p3 -> (r2 -> j3)) -> j3))) -> v3)) -> v3) ->
        (((((k3 -> (u2 ->
79 ((k3 -> (u2 -> t3)) -> t3))) -> ((h3 -> (q3 -> ((h3
        -> (q3 -> d3)) -> d3))) -> y2)) -> y2) -> i3))

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-> i3) -> s2)) -> s2) -> (((((((r4 -> (l4 -> ((
r4 -> (l4 -> z4)) -> z4))) -> ((j4 -> (q4 -> ((
j4 -> (q4 -> f4)) -> f4))) -> k4)) -> k4) ->
80 (((y3 -> (a4 ->
((y3 -> (a4 -> s4)) -> s4))) -> ((h4 -> (g4 -> ((h4
-> (g4 -> e4)) -> e4))) -> b4)) -> b4) -> w4))
-> w4) -> ((((((t4 -> (x4 -> ((t4 -> (x4 -> c4))
-> c4))) -> ((u4 -> (w3 -> ((u4 -> (w3 -> o4))
-> o4))) -> a5)) -> a5) -> (((p4 -> (z3 -> ((p4
-> (z3 -> y
81 4)) -> y4))) -> ((m4 -> (v4 -> ((m4 -> (v4 -> i4))
-> i4))) -> d4)) -> d4) -> n4)) -> n4) -> x3)) ->
x3) -> q2)) -> q2), o2 -> (((((((m3 -> (g3 -> ((
m3 -> (g3 -> u3)) -> u3))) -> ((e3 -> (l3 -> ((e3
-> (l3 -> a3)) -> a3))) -> f3)) -> f3) -> (((t2
-> (v2 -
82 > ((t2 -> (v2 -> n3)) -> n3))) -> ((c3 -> (b3 -> ((
c3 -> (b3 -> z2)) -> z2))) -> w2)) -> w2) -> r3))
-> r3) -> ((((((o3 -> (s3 -> ((o3 -> (s3 -> x2))
-> x2))) -> ((p3 -> (r2 -> ((p3 -> (r2 -> j3))
-> j3))) -> v3)) -> v3) -> (((k3 -> (u2 -> ((k3
-> (u2 ->
83 t3)) -> t3))) -> ((h3 -> (q3 -> ((h3 -> (q3 -> d3))
-> d3))) -> y2)) -> y2) -> i3)) -> i3) -> s2)) ->
s2), e0 -> (h0 -> (i0 -> ((h0 -> (i0 -> j0)) ->
j0))), n2 -> (((((((v1 -> (j1 -> ((v1 -> (j1 ->
k1)) -> k1))) -> ((x1 -> (u1 -> ((x1 -> (u1 -> s1
)) -> s1)
84 )) -> o1)) -> o1) -> (((m1 -> (n1 -> ((m1 -> (n1 ->
p1)) -> p1))) -> ((t1 -> (w1 -> ((t1 -> (w1 ->
q1)) -> q1))) -> r1)) -> r1) -> l1)) -> l1) ->
((((((k2 -> (y1 -> ((k2 -> (y1 -> z1)) -> z1)))
-> ((m2 -> (j2 -> ((m2 -> (j2 -> h2)) -> h2))) ->
d2)) -> d2)
85 -> (((b2 -> (c2 -> ((b2 -> (c2 -> e2)) -> e2))) ->
((i2 -> (l2 -> ((i2 -> (l2 -> f2)) -> f2))) ->
g2)) -> g2) -> a2)) -> a2) -> i1)) -> i1), f0 ->
(k0 -> (l0 -> ((k0 -> (l0 -> m0)) -> m0))), p0
-> (((a1 -> (b1 -> ((a1 -> (b1 -> e1)) -> e1)))
-> ((y0 -> (
86 c1 -> ((y0 -> (c1 -> d1)) -> d1))) -> z0)) -> z0),
n0 -> (((h0 -> (i0 -> ((h0 -> (i0 -> j0)) -> j0))
) -> ((k0 -> (l0 -> ((k0 -> (l0 -> m0)) -> m0)))
-> g0)) -> g0), c0 -> (a0 -> (b0 -> d0)), g1 ->

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      (((((v1 -> (j1 -> ((v1 -> (j1 -> k1)) -> k1))) ->
      ((x1 -> (
87 u1 -> ((x1 -> (u1 -> s1)) -> s1))) -> o1)) -> o1) ->
      (((m1 -> (n1 -> ((m1 -> (n1 -> p1)) -> p1))) ->
      ((t1 -> (w1 -> ((t1 -> (w1 -> q1)) -> q1))) ->
      r1)) -> r1) -> l1)) -> l1), h1 -> (((((k2 -> (y1
      -> ((k2 -> (y1 -> z1)) -> z1))) -> ((m2 -> (j2 ->
88 (j2 -> h2)) -> h2))) -> d2)) -> d2) -> (((b2 -> (c2
      -> ((b2 -> (c2 -> e2)) -> e2))) -> ((i2 -> (l2
      -> ((i2 -> (l2 -> f2)) -> f2))) -> g2)) -> g2) ->
      a2)) -> a2), p2 -> (((((((r4 -> (l4 -> ((r4 -> (
      l4 -> z4)) -> z4))) -> ((j4 -> (q4 -> ((j4 -> (q4
      -> f4))) -> k4)) -> k4) -> (((y3 -> (a4 -> ((y3 ->
89 (a4 -> s4)) -> s4))) -> ((h4 -> (g4 -> ((h4 -> (
      g4 -> e4)) -> e4))) -> b4)) -> b4) -> w4)) -> w4)
      -> (((((((t4 -> (x4 -> ((t4 -> (x4 -> c4)) -> c4)
      )) -> ((u4 -> (w3 -> ((u4 -> (w3 -> o4)) -> o4)))
      -> a5)))
90 -> a5) -> (((p4 -> (z3 -> ((p4 -> (z3 -> y4)) -> y4
      ))) -> ((m4 -> (v4 -> ((m4 -> (v4 -> i4)) -> i4))
      ) -> d4)) -> d4) -> n4)) -> n4) -> x3)) -> x3),
      f1 -> (((((t0 -> (u0 -> ((t0 -> (u0 -> x0)) -> x0)
      ))) -> ((r0 -> (v0 -> ((r0 -> (v0 -> w0)) -> w0))
      ) -> s0)))
91 -> s0) -> (((a1 -> (b1 -> ((a1 -> (b1 -> e1)) ->
      e1))) -> ((y0 -> (c1 -> ((y0 -> (c1 -> d1)) ->
      d1))) -> z0)) -> z0) -> q0)) -> q0))
92 type Int
93 current env is Map(four -> Scheme(Set(s4, q4, y4, d3
      , t4, g3, r2, k3, w2, x3, y2, r3, c3, m4, i4, w3,
      v4, u4, u3, w4, r4, z2, i3, u2, y3, a5, s2, g4,
      f3, t2, n3, l3, v3, c4, f4, x4, x2, h4, j3, t3,
      z4, e3, m3, n4, h3, v2, s3, e4, o4, z3, k4, b4,
      b3, j4, a3
94 , q2, l4, q3, a4, p4, p3, d4, o3),((((((((m3 -> (g3
      -> ((m3 -> (g3 -> u3)) -> u3))) -> ((e3 -> (l3
      -> ((e3 -> (l3 -> a3)) -> a3))) -> f3)) -> f3) ->
      (((t2 -> (v2 -> ((t2 -> (v2 -> n3)) -> n3))) ->
      ((c3 -> (b3 -> ((c3 -> (b3 -> z2)) -> z2))) ->
      w2)) -> w2
95 ) -> r3)) -> r3) -> (((((((o3 -> (s3 -> ((o3 -> (s3
      -> x2)) -> x2))) -> ((p3 -> (r2 -> ((p3 -> (r2 ->

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j3)) -> j3))) -> v3)) -> v3) -> (((k3 -> (u2 ->
((k3 -> (u2 -> t3)) -> t3))) -> ((h3 -> (q3 ->
((h3 -> (q3 -> d3)) -> d3))) -> y2)) -> y2) -> i3
)) -> i3)
96 -> s2)) -> s2) -> (((((((((r4 -> (l4 -> ((r4 -> (l4
-> z4)) -> z4))) -> ((j4 -> (q4 -> ((j4 -> (q4 ->
f4)) -> f4))) -> k4)) -> k4) -> (((y3 -> (a4 ->
((y3 -> (a4 -> s4)) -> s4))) -> ((h4 -> (g4 ->
((h4 -> (g4 -> e4)) -> e4))) -> b4)) -> b4) -> w4
)) -> w4)
97 -> (((((((t4 -> (x4 -> ((t4 -> (x4 -> c4)) -> c4)))
-> ((u4 -> (w3 -> ((u4 -> (w3 -> o4)) -> o4))) ->
a5)) -> a5) -> (((p4 -> (z3 -> ((p4 -> (z3 ->
y4)) -> y4))) -> ((m4 -> (v4 -> ((m4 -> (v4 -> i4
)) -> i4))) -> d4)) -> d4) -> n4)) -> n4) -> x3))
-> x3) ->
98 q2)) -> q2)), three -> Scheme(Set(j2, i1, m1, c2,
n1, r1, z1, g2, q1, s1, w1, t1, f2, x1, o1, j1,
i2, a2, h2, b2, u1, v1, p1, k2, m2, l2, l1, y1,
e2, k1, d2),((((((v1 -> (j1 -> ((v1 -> (j1 ->
k1)) -> k1))) -> ((x1 -> (u1 -> ((x1 -> (u1 ->
s1)) -> s1))) -
99 > o1)) -> o1) -> (((m1 -> (n1 -> ((m1 -> (n1 -> p1)
) -> p1))) -> ((t1 -> (w1 -> ((t1 -> (w1 -> q1))
-> q1))) -> r1)) -> r1) -> l1)) -> l1) -> ((((((
k2 -> (y1 -> ((k2 -> (y1 -> z1)) -> z1))) -> ((m2
-> (j2 -> ((m2 -> (j2 -> h2)) -> h2))) -> d2))
-> d2) ->
100 (((((b2 -> (c2 -> ((b2 -> (c2 -> e2)) -> e2))) -> ((
i2 -> (l2 -> ((i2 -> (l2 -> f2)) -> f2))) -> g2))
-> g2) -> a2)) -> a2) -> i1)) -> i1)), two ->
Scheme(Set(u0, x0, q0, a1, b1, s0, e1, d1, z0, w0
, y0, v0, t0, c1, r0),((((t0 -> (u0 -> ((t0 -> (
u0 -> x0))
101 -> x0))) -> ((r0 -> (v0 -> ((r0 -> (v0 -> w0)) -> w0
))) -> s0)) -> s0) -> (((a1 -> (b1 -> ((a1 -> (
b1 -> e1)) -> e1))) -> ((y0 -> (c1 -> ((y0 -> (c1
-> d1)) -> d1))) -> z0)) -> z0) -> q0)) -> q0)),
main -> Scheme(Set(),Int), tuple -> Scheme(Set(
a0, b0, d0
102 ),(a0 -> (b0 -> ((a0 -> (b0 -> d0)) -> d0))), one
-> Scheme(Set(k0, g0, h0, i0, l0, m0, j0),(((h0
-> (i0 -> ((h0 -> (i0 -> j0)) -> j0))) -> ((k0 ->
(l0 -> ((k0 -> (l0 -> m0)) -> m0))) -> g0)) ->

```

```
    g0)))  
103 Type vars = 132  
104 Type vars = 132  
105 Type vars in sub = 472  
106 ##### main #####
```

$$\begin{aligned}
& \lambda f_1. \lambda f_2. \lambda f_3. (f_1 f_2) f_3 & (5.1) \\
& = \lambda f_1. \lambda f_2. \sigma(\lambda f_3. f_1 f_2)(\lambda f_3. f_3) \\
& = \lambda f_1. \lambda f_2. \sigma(\sigma(\lambda f_3. f_1)(\lambda f_3. f_2))(\iota) \\
& = \lambda f_1. \lambda f_2. (\sigma(\sigma(\kappa f_1)(\kappa f_2)))\iota \\
& = \lambda f_1. \sigma(\lambda f_2. \sigma(\sigma(\kappa f_1)(\kappa f_2)))(\lambda f_2. \iota) \\
& = \lambda f_1. \sigma(\sigma(\lambda f_2. \sigma)(\lambda f_2. (\sigma(\kappa f_1)(\kappa f_2)))(\lambda f_2. \iota) \\
& = \lambda f_1. \sigma(\sigma(\kappa \sigma)(\sigma(\lambda f_2. \sigma(\kappa f_1))(\lambda f_2. \kappa f_2)))(\lambda f_2. \iota) \\
& = \lambda f_1. \sigma(\sigma(\kappa \sigma)(\sigma(\sigma(\lambda f_2. \sigma)(\lambda f_2. \kappa f_1))(\sigma(\lambda f_2. \kappa)(\lambda f_2. f_2))))(\kappa \iota) \\
& = \lambda f_1. \sigma(\sigma(\kappa \sigma)(\sigma(\sigma(\kappa \sigma)(\sigma(\lambda f_2. \kappa)(\lambda f_2. f_1)))(\sigma(\kappa \kappa)(\iota))))(\kappa \iota) \\
& = \lambda f_1. \sigma(\sigma(\kappa \sigma)(\sigma(\sigma(\kappa \sigma)(\sigma(\kappa \kappa)(\kappa f_1)))(\sigma(\kappa \kappa)(\iota))))(\kappa \iota) \\
& = \lambda f_1. (\sigma(\sigma(\kappa \sigma)(\sigma(\sigma(\kappa \sigma)(\sigma(\kappa \kappa)(\kappa f_1)))(\sigma(\kappa \kappa)(\iota)))))(\kappa \iota) \\
& = \sigma((\lambda f_1. \sigma)(\lambda f_1. (\sigma(\kappa \sigma)(\sigma(\sigma(\kappa \sigma)(\sigma(\kappa \kappa)(\kappa f_1)))(\sigma(\kappa \kappa)(\iota)))))(\lambda f_1. \kappa \iota) \\
& = \sigma((\kappa \sigma)(\sigma(\lambda f_1. \sigma(\kappa \sigma))(\lambda f_1. (\sigma(\sigma(\kappa \sigma)(\sigma(\kappa \kappa)(\kappa f_1)))(\sigma(\kappa \kappa)(\iota)))))(\sigma(\lambda f_1. \kappa)(\lambda f_1. \iota)) \\
& = \sigma((\kappa \sigma)(\sigma((\lambda f_1. \sigma)(\lambda f_1. \kappa \sigma))(\sigma(\lambda f_1. \sigma(\sigma(\kappa \sigma)(\sigma(\kappa \kappa)(\kappa f_1)))(\lambda f_1. (\sigma(\kappa \kappa)(\iota)))))(\sigma(\kappa \kappa)(\kappa \iota)) \\
& = \sigma \\
& \quad ((\kappa \sigma)(\sigma((\kappa \sigma)(\sigma(\lambda f_1. \kappa)(\lambda f_1. \sigma)))(\sigma(\sigma(\lambda f_1. \sigma)(\lambda f_1. (\sigma(\kappa \sigma)(\sigma(\kappa \kappa)(\kappa f_1)))(\sigma(\lambda f_1. \sigma(\kappa \kappa)(\lambda f_1. \iota)))))) \\
& \quad (\sigma(\kappa \kappa)(\kappa \iota)) \\
& = \sigma \\
& \quad ((\kappa \sigma)(\sigma((\kappa \sigma)(\sigma(\kappa \kappa)(\kappa \sigma)))(\sigma(\sigma(\kappa \sigma)(\sigma(\lambda f_1. \sigma(\kappa \sigma)))(\lambda f_1. (\sigma(\kappa \kappa)(\kappa f_1)))(\sigma(\sigma(\kappa \sigma)((\kappa \kappa)(\kappa \kappa)))(\kappa \iota)))))) \\
& \quad (\sigma(\kappa \kappa)(\kappa \iota)) \\
& = \sigma \\
& \quad ((\kappa \sigma)(\sigma((\kappa \sigma)(\sigma(\kappa \kappa)(\kappa \sigma)))(\sigma(\sigma(\kappa \sigma)(\sigma(\sigma(\lambda f_1. \sigma)(\lambda f_1. \kappa \sigma)) \\
& \quad (\sigma(\lambda f_1. \sigma(\kappa \kappa)))(\lambda f_1. \kappa f_1)))))(\sigma(\sigma(\kappa \sigma)((\kappa \kappa)(\kappa \kappa)))(\kappa \iota)))) \\
& \quad (\sigma(\kappa \kappa)(\kappa \iota)) \\
& = \sigma \\
& \quad ((\kappa \sigma)(\sigma((\kappa \sigma)(\sigma(\kappa \kappa)(\kappa \sigma)))(\sigma(\sigma(\kappa \sigma)(\sigma(\sigma(\kappa \sigma)((\kappa \kappa)(\kappa \sigma)) \\
& \quad (\sigma(\sigma(\lambda f_1. \sigma)(\lambda f_1. \kappa \kappa)))(\sigma(\lambda f_1. \kappa)(\lambda f_1. f_1)))))(\sigma(\sigma(\kappa \sigma)((\kappa \kappa)(\kappa \kappa)))(\kappa \iota)))) \\
& \quad (\sigma(\kappa \kappa)(\kappa \iota)) \\
& = \sigma \\
& \quad ((\kappa \sigma)(\sigma((\kappa \sigma)(\sigma(\kappa \kappa)(\kappa \sigma)))(\sigma(\sigma(\kappa \sigma)(\sigma(\sigma(\kappa \sigma)((\kappa \kappa)(\kappa \sigma)) \\
& \quad (\sigma(\sigma(\kappa \sigma)((\kappa \kappa)(\kappa \kappa)))(\sigma(\kappa \kappa)(\iota)))))(\sigma(\sigma(\kappa \sigma)((\kappa \kappa)(\kappa \kappa)))(\kappa \iota)))) \\
& \quad (\sigma(\kappa \kappa)(\kappa \iota)) \\
& = S \\
& \quad ((KS)(S((KS)(S(KK)(KS)))(S(S(KS)(S(S(KS)((KK)(KS))) \\
& \quad (S(S(KS)((KK)(KK)))(S(KK)(I)))))(S(S(KS)((KK)(KK)))(KI)))) \\
& \quad (S(KK)(KI))
\end{aligned}$$

