Interactive Type Debugging in Haskell

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ABSTRACT
In this paper we illustrate the facilities for type debugging of Haskell programs in the Chameleon programming environment. Chameleon provides an extension to Haskell supporting advanced and programmable type extensions. Chameleon maps the typing problem for a program to a system of constraints each attached to program code that generates the constraints. We use reasoning about constraint satisfiability and implication to find minimal justifications of type errors, and to explain unexpected types that arise. Through an interactive process akin to declarative debugging, a user can track down exactly where a type error occurs. The approach handles Hindley/Milner types with Hindley/Milner types with Haskell-style overloading. The Chameleon system provides a full implementation of our flexible type debugging scheme which can be used as a front-end to any existing Haskell system.

Categories and Subject Descriptors
D.3.2 [Programming Languages]: Language Classifications—Applicative (functional) languages; D.3.3 [Programming Languages]: Language Constructs and Features—Polymorphism, Constraints; F.3.3 [Logics and Meanings of Programs]: Studies of Program Constructs—Type structure

General Terms
Languages, Theory

Keywords
type inference, type debugging, Hindley/Milner, overloading, type classes, constraints

1. INTRODUCTION
Strongly typed languages provide the user with the convenience to significantly reduce the number of errors in a program. Well-typed programs can be guaranteed not to "go wrong" [21], with respect to a large number of potential problems. However, programs are often not well-typed, and therefore must be modified before they can be accepted. Unfortunately, it can be difficult to determine why a program has been rejected.

Traditional inference algorithms depend on a particular traversal of the syntax tree. Therefore, inference frequently reports errors at locations which are far away from the actual source of the problem. The programmer is forced to tackle the problem of correcting her program unaided. This can be a daunting task for even experienced programmers; beginners are often left bewildered.

Despite recent efforts, see e.g. [3, 20], we believe there remains a lot of scope for improvement. For example, previous works exclude Haskell-style overloading [27] which can be naturally handled by our approach (see e.g. Examples 2 and 7), and is a serious cause for consternation for beginning Haskell programmers.

The novelty of our approach lies in mapping the entire typing problem, including type classes and extensions, to a set of constraints. Program locations are attached to individual constraints. By employing some simple constraint reasoning steps we are able to narrow down the source of the type error. We demonstrate our approach via a simple example.

Example 1. Consider the following annotated program where we use numbers to refer to individual program locations.

\[ p_4 = (f_2 \ 'a'_1)_3 \]
\[ f_7 \ True_5 = True_6 \]

Each expression is translated into a set of constraints, each of which has an attached justification in this case a program location number, and a type variable representing its type. For example, expression \( f_2 \ 'a'_1 \) is translated to constraint \( t_1 = \text{Char} \), \( f(t_2)_2, t_2 = t_1 \rightarrow t_3 \) with type \( t_3 \). Note that we introduce for each function symbol \( f \) a predicate symbol \( f \). So, for example \( f(t_2)_2 \) refers to an instance of function \( f \) at location 2 where \( t_2 \) refers to the particular instance type.

Each function definition is translated to a Constraint Handling Rule (CHR) [6]. More specifically, we make use of CHR simplification rules. For example, in case of the above
program we find
\[
p(t_8) \iff (t_1 = \text{Char}(1,8), f(t_2) = (t_2 = t_1 \rightarrow t_3)_{1,8}, t_4 = t_3)_{1,8}
\]
\[
f(t_7) \iff (t_5 = \text{Bool}, t_6 = \text{Bool} = t_6, t_7 = t_6 \rightarrow t_0)_{1,7}
\]
The left-hand side of the first rule consists of the predicate \(p(t_8)\) where \(t_8\) refers to the type of function \(p\). The right-hand side consists of the constraints generated out of the program text representing the type of \(p\). The \(\iff\) symbol can be read as logical equivalence. Operationally the rules are read as defining replacements, you may replace the left hand side by the right hand side. We keep applying rules until no further rules are applicable.

For example, we can infer the type of expression \(p_8\) by applying the above CHRs to the initial constraints store \(p(t_8)\) where \(8\) stands for a hypothetical program location. Here is the final result.

\[
p(t_8) \iff (t_1 = \text{Char}(1,8), f(t_2) = (t_2 = t_1 \rightarrow t_3)_{1,8}, t_4 = t_3)_{1,8}
\]
\[
= t_7 = t_5 \rightarrow t_0)_{1,7}
\]

In the first step, the constraint \(p(t_8)\) matches the left hand side of the first CHR. We replace \(p(t_8)\) by the right hand side. In addition, we add the matching equation \(t = t_4\). Note how the justification from \(p(t_8)\) is added to each justification set. In the final step, the constraint \(f(t_2) = (t_2 = t_1 \rightarrow t_3)_{1,8}\) matches the left hand side of the second CHR.

Note that constraints are unsatisfiable in the final store. Indeed, \(p\) is not well-typed. By collecting justifications attached to unsatisfiable constraints we are able to narrow down the possible source of the type error. A minimal unsatisfiable subset of the resulting constraints is \((t_1 = \text{Char}(1,8), t_2 = t_7, t_5 = \text{Bool})_{1,8}, (t_7 = t_5 \rightarrow t_0)_{1,7}\). Hence the system underlines the program location \{1,2,3,5,7\} (ignoring 8 since we did not provide it in the program).

\[
p = f \cdot 'a'
\]
\[
\text{if } \text{True} = \text{True}
\]
indicating that the program must be changed in at least one of these locations to be corrected.

Haack and Wells [9] in parallel proposed a very similar approach to that above, mapping the typing problem to Herbrand constraints. The advantage of using CHRs arises when we extend the approach to handle Haskell-style overloading [27].

**Example 2.** Consider the following program making use of Haskell type class \(\text{Ord}\) via the \(<\) function.

\[
q \ x \ y \ z = \text{if } x < y \text{ then } z \text{ else } y \ z
\]

There is a type error since we are comparing a function \(y\). The Hugs systems reports

```
ERROR "h.hs" (line 2): Illegal Haskell 98 class constraint in inferred type
*** Expression : q
*** Type : \text{Ord} \rightarrow \rightarrow \rightarrow \rightarrow \rightarrow\rightarrow
```

In our system we can ask for an explanation for why the offending class constraint appears. Instead of minimal unsatisfiable subsets we simply search for minimal implicants.

Justifications attached to minimal implicants allows us to narrow down the possible source of the unexpected type. In Chameleon, we say \(\text{explain} q \ (\text{Ord} \rightarrow \rightarrow \rightarrow)\) (where \(\ast\) stands for an anonymous type variable) and the system reports

\[
q \ x \ y \ z = \text{if } x < y \text{ then } z \text{ else } y \ z
\]

illustrating why the offending constraint arises.

Since many Haskell Prelude functions make use of type class overloading, the ability to handle this feature is vital for a Haskell type debugger.

Our type inference algorithm generates equational and user-defined constraints out of expressions. Constraints are justified by the program location where they originated from. These locations are retained during CHR solving. Simple reasoning steps on constraints, such as finding minimal unsatisfiable subsets and minimal implicants, allows us to identify problematic program locations.

Our contributions are:

- We give a translation of the typing problem for Hindley/Milner which includes Haskell-style overloading into CHRs.
- We refine CHR solving by keeping track of justifications, i.e. program locations, attached to constraints.
- Our approach is the first to:
  - explain the locations that lead to a type having a certain shape,
  - handle Haskell-style overloading (and indeed more complex type extensions [24]).
- We provide an interactive type debugger implementing the above ideas as part of the Chameleon environment.

The rest of the paper is organized as follows. We first describe the debugging features supported by the Chameleon debugging system, with only informal explanations about how they are implemented. Then we give the formal underpinning to the system. In Section 3 where we introduce types and constraints and then present a formal definition of constraint solving in terms of Constraint Handling Rules (CHRs) in Section 4. In Section 5 we show how to translate a Hindley/Milner typing problem with Haskell-style overloading into a system of CHRs, and how we use this for type inference and checking in Section 6. In Section 7, we discuss how simple constraint reasoning steps support type debugging of programs. Related work is discussed in Section 8. We conclude in Section 9. An implementation of our type debugger is available via [26].

## 2. The Chameleon Type Debugger

In this section we explain the features of the Chameleon type debugger. Chameleon can be used as a front-end to any existing Haskell system. The Chameleon system does not currently allow for the debugging of errors in the definitions of type classes and instances.\(^1\) Chameleon does of course allow us to debug the usage of classes and instances.

\(^1\)To do so requires a well-understood check of the confluence of the CHRs [1]. This is straightforward for CHRs arising from Haskell 98 classes and instances, but termination issues arise when arbitrary programmed type extensions are allowed. Currently, we also do not check for the monomorphism restriction and some other Haskell 98 specific context restrictions.
The debugger makes use of two kinds of constraint reasoning. In order to explain why a type error arises it determines minimal unsatisfiable subsets of a set of constraints. In order to explain why an expression has a certain type it determines a minimal implicant of a set of constraints. Details of these operations can be found in Section 7.

2.1 Error Explanation

We can ask for the type of an expression e using the command type e. If e has no type this displays the parts of the program which cause the error. It translates e into a set of constraints C. The constraint C is executed with respect to the CHRs P of the translated program, by exhaustively applying the rules in P to obtain a new constraint C'. We denote this execution by C →* C'. If C' is satisfiable it displays the type of e. Otherwise the system determines a minimal unsatisfiable subset of C' (simply the first one detected) and displays the justifications for that set.

We support two basic approaches to displaying the justifications of an error.

Local explanation restricts attention to the expression for which the type error is detected, all locations outside this expression are ignored. If the expression is a single function name, we restrict attention to the function definition. Local explanation is useful for top-down exploratory style of type debugging which we believe is more natural. Indeed while using local explanations the system in fact simplifies all constraints arising from each other function, which considerably simplifies the calculation of minimal unsatisfiable subsets and minimal implicants.

**Example 3.** Returning to Example 1, using local explanation the CHR for \( f(t) \) is treated as \( f(t) \iff t = \text{Bool} \rightarrow \text{Bool} \) and the minimal unsatisfiable subset is \((t_1 = \text{Char} \rightarrow 1, 8), (t_2 = \text{Bool} \rightarrow \text{Bool}) \rightarrow 2, 8, (t_2 = t_1 \rightarrow t_3) \rightarrow 3, 8\). The resulting justification 1, 2, and 3 is the same as before (restricted to locations in p), but less constraints are generated. The resulting explanation is \( p = f^{2,3,8} \).

Global explanation is more expensive to compute but allows the user to explore an error in a bottom-up manner. Here all the justification information is used to determine an error. The system highlights positions involved in one minimal unsatisfiable subset, and can highlight those positions that occur in all minimal unsatisfiable subsets differently from those not occurring in all.

**Example 4.** Consider the following program:

\[
\begin{align*}
\text{foldl } f z \ [z] & = [z] \\
\text{foldl } f z (x:xs) & = \text{foldl } f (f z x) xs \\
\text{flip } f x y & = f y x \\
\text{reverse } & = \text{foldl} (\text{flip} ;) [] \\
palin xs & = \text{reverse } xs = xs
\end{align*}
\]

where palin is intended to be a function from a list to a Boolean value. It should return True if its argument is a palindrome. Hugs [15] reports the following:

```
ERROR Ex1.hs:6 - Type error in application
*** Expression : xs == reverse xs
*** Term : xs
*** Type : [a]
*** Does not match : [[a]]
*** Because : unification would give infinite type
```

By using global error explanation, we can get an immediate picture of the program sites which interact to cause this error. In the following debugger query, one global explanation of the type error is given by the underlined code. Locations which appear in all minimal unsatisfiable subsets are underlined twice, while those which only appear in the selected minimal unsatisfiable subset are underlined once.

Note that the "real" cause of the error does occur in all unsatisfiable subsets, and in our experience this is usually the case where there is one "real" error.

```
Ex1.hs> :set global
Ex1.hs> :type palin
palin :: [a] -> Bool
```

```
foldl f z [] = [z]
foldl f z (x:xs) = foldl f (f z x) xs
flip f x y = f y x
reverse = foldl (flip ;) []
palin xs = reverse xs = xs
```

By starting from any of these sites, the programmer is able to work towards the true cause of the error - in any direction. In this case, the problem is that the first clause of foldl should not return a list, [z], but rather z. Correspondingly, the offending location is double-underlined.

If we had taken a local approach to error explanation here, the result would have been specific to only the definition of palin. We would then have to proceed in a top-down fashion, from definition to definition, towards the offending expression.

The system naturally handles type class extensions that can be expressed by CHRs.

**Example 5.** Functional dependencies in class constraints are useful for preventing ambiguity. Consider a multi-parameter collection class `Collect a b` where type `b` is a collection of elements of type `a`. The class definition is

```
class Collect a b where
  empty :: b
  insert :: a -> b -> b
  member :: a -> b -> Bool
```

As defined this class is flawed, since the type of `empty :: Collect a b -> b` is ambiguous. Type variable `a` appears only in the constraint component. This leads to difficulties when implementing Haskell-style overloading [16]. Functional dependencies allow us to overcome this problem, by stating that `b` functionally determines `a`. Hugs [15] supports functional dependencies.

```
class Collect a b | b -> a where ...
```

This functional dependency can be expressed by a CHR propagation rule.

```
Collect a b, Collect a' b == a = a'
```

The \( \Rightarrow \) symbol is read as logical implication. Operationally the rule is read as, if you have a match for the left hand side you may add the right hand side. The above rule states that if there are two `Collect` constraints with the same second argument, we enforce that their first arguments are identical.

Consider the following program which tries to check if a `Float` is a member of a collection of `Ints`.

```
coll :: [Int] -> [Float] -> Bool
```
The inferred type has a class constraint that is non-variable and has no instance. This is completely opaque to a beginning Haskell programmer.

We can generate an explanation for the error by looking for a reason for (e.g. minimal set of constraints implying) the constraint Num [a]. Asking the type debugger

```
Ex11.hs> :explain sum (Num [a] => *)
sum [] = []
sum (x:xs) = x + sum xs
```

Clearly indicating the problem arises from the [a] of the body of the first rule interacting with + and the recursive call to sum.

**Example 8. Returning to Example 5, the Hugs error message is**

```
ERROR Ex9.hs:10 - Constraints are not consistent with functional dependency
*** Constraint : Collects Float a
*** And constraint : Collects Int a
*** For class : Collects a b
*** Break dependency : b -> a
```

This gives very little information to the programmer about the error. In our system we can ask where the constraints arise from:

```
Ex9.hs> :explain f (Collects Float * => *)
f g x y = if member (x::Float) (insert (1::Int) y)
then g x else y
```

Note that even though the constraint system is unsatisfiable, a minimal implicant correctly determines a useful justification of the constraint.

### 2.3 Referring to Local Variables

In order to track down type errors interactively it is often useful to be able to find the types and explain the types of variables local to a function definition. Current interactive Haskell systems only permit references to variables bound at the top-level. The debugger allows the syntax f;r to refer to variable r local to the definition of f. If there are multiple equations (patterns) for the definition for f we can select one using the notation f;n;r where n is an integer pattern number. By default if there are multiple equations, and no pattern number is given, the first where the local variable exists is used. Local variables inside nested scopes can also be referred to.

**Example 9. Consider the program**

```
f (x:xs1) _ = True
f xs2 ys = let h xs; = xs ++ xs
          g ys = ys ++ xs
          in h ys ++ g (xs ++ ys)
```

Then f;1;xs refers to xs1, while f;2;zs refers to xs2. By default f;zs refers to xs1. In addition f;2;h;zs and f;h;zs refer to zs.

### 2.4 Type Addition

While the explain command allows users to ask why a location has a type of a particular shape, the declare command allows users to ask why not of a location and a type. The declare f (C => t) command adds constraints x = t, C where x is the type of f to the CHR program defining f.

```
```
EXAMPLE 10. Returning to Example 7, we can get another explanation for the erroneous type of sum by adding the expected type.

Ex11.hs> :declare sum ([Int] -> Int)
Ex11.hs> :type sum
sum :: [Int] -> Int

we are shown those locations which are in conflict with the newly declared type.

2.5 Source-Based Debugger Interface

Although an interactive debugging system provides users with the means to quickly pose a number of consecutive queries, narrowing in on the target, it might also be viewed as a slightly heavy handed interface to the debugger. An interactive system necessarily interrupts the typical, edit-compile programming cycle, which may be distracting. Furthermore, it may at times seem quite awkward to keep track exploration separate from the program source itself.

To this end we have provided an alternative means to interact with the debugger, by allowing for commands to appear naturally within the source program. At this time we have support for type e where e is an expression written e::?. And we support explain e (D ⇒ t), where e is an expression and D ⇒ t is a type scheme, written e::?:D⇒.

Entire declarations can be queried by writing such a command at the same scope as the declaration (with a declaration name in place of an expression.) These queries are collected and processed in textual order. They do not effect the semantics of the program in which they are embedded, merely the compiler’s output.

EXAMPLE 11. Consider the following, modified, snippet of the program presented in Example 4.

reverse ::?
reverse = foldl (flip ())) []

When we attempt to compile this code, using the non-interactive system, we would get, in addition to the usual type error message, the following output:

reverse :: [a] ⇒ [[a]]

Further modification of the program might lead to the following, which involves an explain-style query:

reverse ::?
reverse = foldl (flip ())) [] :: [+] ⇒ [[+]]

The corresponding output would be:

reverse :: [a] ⇒ [[a]]

foldl (flip ())) [] :: [+] ⇒ [[+]]
because of: foldl (flip ())) []

2.6 Declarative Debugging Interface

Chameleon also includes a simplistic declarative debugging interface. We can invoke the declarative debugging interface on an expression e with a type error using the command debug e. The declarative debugger works like a declarative debugger for a logic program [23], localizing the error by finding a function whose type is wrong, but all the functions used in its definitions are correct. A similar feature is also provided by [14, 3].

EXAMPLE 12. Consider the program of Example 4 once more. The declarative debugger trace might be

Ex1.hs> :debug palin
reverse :: [a] ⇒ [[a]]
Ex1.hs> is this type correct> n
flip :: (a ⇒ b ⇒ c) ⇒ b ⇒ a ⇒ c
Ex1.hs> is this type correct> y
foldl :: (a ⇒ b ⇒ a) ⇒ a ⇒ [b] ⇒ [a]
Ex1.hs> is this type correct> y

The declarative debugging interface, chooses a minimal unsatisfiable subset, and asks the user about types of functions involved in this set, from the top down to discover where the error actually lies. It then shows the justifications of the error in this function. Note that since it uses a minimal unsatisfiable subset, it will never ask questions about functions not involved in this subset. This is not the case for the system of [3], since it does not determine minimal unsatisfiable subsets.

3. TYPES AND CONSTRAINTS

We consider an extension of the Hindley/Milner system with constraints.

Expressions e ::= f | x | λx.e | e | let f = e in e
Types t ::= a | t → t | T ·
Type Schemes σ ::= τ | ∀a.ρ ⇒ t
Constraints C ::= t = t | U t | C ∧ C

...
the equation \( \theta \sigma \) suppose there exists substitution \( E \) given by of
that \( \{ \) general unifier of \( c \) where
hand side of rules have attached justifications (program lo-
is defined as follows. Let \( \{ \)
be such that the most
logical entailment relation and \( \triangleright \) stands for logical implication. We generally assume that \( F \) can be described by a set of CHRs.

\[ \Gamma \vdash \varnothing \]

\[ \forall \]

4. CONSTRAINT HANDLING RULES WITH JUSTIFICATIONS

We will translate the typing problem to a constraint problem
where the meaning of the user-defined constraints is de-
defined by Constraint Handling Rules (CHR) [6]. CHRs ma-
nipulate a global set of primitive constraints, using rewrite rules of two forms

\[ \text{simplification} \quad c_1, \ldots, c_n \iff d_1, \ldots, d_m \]

\[ \text{propagation} \quad c_1, \ldots, c_n \implies d_1, \ldots, d_m \]

where \( c_1, \ldots, c_n \) are user-defined constraints, and \( d_1, \ldots, d_m \) are constraints.

The logical interpretation of the rules is as follows. Let \( \bar{x} \) be the variables occurring in \( \{ c_1, \ldots, c_n \} \), and \( \bar{y} \) be the other variables occurring in the rule. The logical reading is

\[ \text{simplification} \quad \forall \bar{x}( (c_1 \land \cdots \land c_n) \iff \exists \bar{y} (d_1 \land \cdots \land d_m) ) \]

\[ \text{propagation} \quad \forall \bar{x}( (c_1 \land \cdots \land c_n) \implies \exists \bar{y} (d_1 \land \cdots \land d_m) ) \]

In our use of the rules, constraints occurring on the right hand side of rules have attached justifications (program locations). We extend the usual derivation steps of Constraint Handling Rules to maintain justifications.

A simplification derivation step applying a renamed rule instance \( r \equiv c_1, \ldots, c_n \implies d_1, \ldots, d_m \) to a set of constraints \( C \) is defined as follows. Let \( E \subseteq C_U \) be such that the most general unifier of \( E \) is \( \theta \). Let \( D = \{ c'_1, \ldots, c'_m \} \subseteq C_U \) and suppose there exists substitution \( \sigma \) on variables in \( r \) such that \( \theta(c'_1), \ldots, \theta(c'_m) \vdash \sigma(c_1), \ldots, \sigma(c_n) \), that is a subset of \( C_U \) matches the left hand side of \( r \) under the substitution given by \( E \). The justification \( J \) of the matching is the union of the justifications of \( E \cup D \).

Then we create a new set of constraints \( C' = C - \{ c'_1, \ldots, c'_m \} \cup \{ \theta(c'_1) = c_1, \ldots, \theta(c'_m) = c_n, (d_1)_{+J}, \ldots, (d_m)_{+J} \} \). Note that the equation \( \theta(c'_i) = c_i \) is shorthand for \( \theta(s_1) = t_1, \ldots, \theta(s_m) = t_m \) where \( c'_i \equiv p(s_1, \ldots, s_m), p \) and \( c_i \equiv p(t_1, \ldots, t_m) \).

A derivation step from global set of constraints \( C \) to \( C' \) using an instance of rule \( r \) is denoted \( C \rightarrow_r C' \). A derivation, denoted \( C \rightarrow^* C' \), is a sequence of derivation steps using rules in \( P \) where no derivation step is applicable to \( C' \). The operational semantics of CHRs exhaustively apply rules to the global set of constraints, being careful not to apply propagation rules twice on the same constraints (to avoid infinite propagation). For more details on avoiding rep propagation see e.g. [1].

5. TRANSLATION TO CONSTRAINT HANDLING RULES

Our approach to type inference follows [4] by translating the typing problem into a constraint problem. However, in contrast to [4] where translation results in a set of Horn clauses, we map the typing problem to a set of Constraint Handling Rules (CHR) [6].

For each definition \( f = e \), we introduce a CHR of the form \( f(t, l) \iff C \). The type parameter \( l \) refers to the type of \( f \) whereas \( l \) refers to the set of types of \( \lambda \)-bound variables in scope (i.e. the set of types of free variables which come from the enclosing definition). The reason for \( l \) is that we must ensure that \( \lambda \)-bound variables remain monomorphic. The constraint \( C \) contains the constraints generated out of expression \( e \) plus some additional constraints restricting \( l \). We use list notation (on the level of types) to refer to the “set” of types of \( \lambda \)-bound variables. In order to avoid con

\[ k z = \text{let}~ h w = (w, z) \]

\[ f x = \text{let}~ g y = (x, y) \]
A (partial) description of the resulting CHRs might look as follows. For simplicity, we leave out the constraints generated out of expressions. We commonly write \( t_z \) to denote the type of \( \lambda \)-bound variable \( z \).

\[
\begin{align*}
(k) & \quad k(t,l) \iff l = \langle \rangle, \\
(h) & \quad h(t,l) \iff l = \langle t_z \rangle, \\
(f) & \quad f(t,l) \iff l = \langle t_z \rangle, \\
(g) & \quad g(t,l) \iff l = \langle t_z, t_w \rangle.
\end{align*}
\]

Note that the \( \lambda \) parameter \( l \) refers exactly to the set of types of all free (\( \lambda \)) variables in scope.

Consider expression \((g_1, g \text{ True}, h \text 3)\). At each instantiation site, we need to specify correctly the set of types of \( \lambda \)-bound variables in scope which were in scope at the function definition site. Note that \( \lambda \)-variables \( z \) and \( x \) are in scope of \( g y \ldots \) whereas only \( z \) is in scope of \( h w \ldots \). Among others, we generate (justifications are omitted for simplicity)

\[
\begin{align*}
g(t_1, t_1), l_1 &= \langle t_z, t_w \rangle, t_1 = \text{Int} \rightarrow t'_1, \\
g(t_2, t_2), l_2 &= \langle t_z, t_w \rangle, t_2 = \text{Bool} \rightarrow t'_2, \\
h(t_3, t_3), l_3 &= \langle t_z, t_w \rangle, t_3 = \text{Int} \rightarrow t'_3.
\end{align*}
\]

We observe that at function instantiation sites our constraint generation algorithm needs to remember correctly the set of types of \( \lambda \)-variables where were in scope at the function definition site. We apply a trick to avoid such calculations. The set of types of \( \lambda \)-bound variables in scope for function definitions is left “open”. The set of types of \( \lambda \)-bound variables at function instantiation sites corresponds to the “full” set of types of \( \lambda \)-bound variables in scope. Our actual translation yields the following result.

\[
\begin{align*}
(k) & \quad k(t,l) \iff l = r, t = t_1 \rightarrow t_2, f(t, l_1), l_1 = \langle t_z \rangle, t_1 = t_z \\
(h) & \quad h(t,l) \iff l = \langle t_z, r \rangle, t = t_w \rightarrow \langle t_w, t_z \rangle \\
(f) & \quad f(t,l) \iff l = \langle t_z, r \rangle, t = (t_1, t_2, t'_2), g(t_1, l_1), \\
& \quad t_1 = \langle t_z, t_w \rangle, t = \text{Int} \rightarrow t'_1, \\
& \quad g(t_2, t_2), t_2 = \langle t_z, t_w \rangle, t_2 = \text{Bool} \rightarrow t'_2, \\
& \quad h(t_3, t_3), l_3 = \langle t_z, t_w \rangle, t_3 = \text{Int} \rightarrow t'_3 \\
(g) & \quad g(t,l) \iff l = \langle t_z, t_2, r \rangle, t = t_y \rightarrow \langle t_z, t_y \rangle.
\end{align*}
\]

For example, in rule \((h)\), we need to check that \( z \) is in scope plus possibly some more variables (see underlined constraint). Please observe that in rule \((f)\), we pass in the (somewhat redundant) variable \( t_x \) as part of the \( l \) parameter at the instantiation site of \( h \) (see double-underlined constraint). There is no harm in doing so, because there is no reference to \( t_x \) on the right hand side of rule \((k)\).

The translation of the typing problem consists of a mutually recursive process of generating constraints out of expressions and generating CHRs for function definitions. We assume that individual expressions are annotated with unique numbers, i.e. program locations.

Constraint generation is formulated as a logical deduction system with clauses of the form \( \Gamma, e \vdash_{\text{Cons}} (C, I) \). For example, in rule \((\text{Var-f})\) we generate an “instantiation” constraint. The constraint \( f(l, l), l = \langle t_x, t_y, \ldots, t_m \rangle \) demands on instance of \( f \) on type \( t \) such that \( (t_x, t_y, \ldots, t_m) \) refers to the set of types of \( \lambda \)-bound variables in scope. The actual type of \( f \) will be described by a CHR where the set of types of \( \lambda \)-bound variables is left open. Note that the order of types of lambda-bound variables matters.

Generation of CHRs is formulated as a logical deduction system with clauses of the form \( \Gamma, e \vdash_{\text{Cons}} P \). For simplicity, we leave out the constraints generated out of expressions. We assume that individual expressions are annotated with unique numbers, i.e. program locations.

5.1 Type Annotations

A type annotation \( \langle f :: C \rightarrow t \rangle \) generates the CHR \( f_\Gamma(t) \iff (t' = \langle t_1 \rangle, C) \). The CHR generated for \( g \) is

\[
\begin{align*}
g_\Gamma(t) & \iff (t_2 = t_z), (t_3 = a \rightarrow [a] \rightarrow [a]), \\
& \quad (t_4 = t_2 \rightarrow t_4), (t_5 = t_8), (t_6 = t_8 \rightarrow t_8), (t_7 = \text{Bool})_t, (t_8 = \text{Bool})_r, g_\Gamma(t_8)
\end{align*}
\]

For functions with both a definition and an annotation we additionally have to check that the annotated type is “subsumed” by the inferred type. Details will be discussed in Section 6.

5.2 Multiple Clauses

The (possibly multiple) definitions for a single function are joined using another CHR. If there are \( m \) definitions of \( f \), numbered \( f_1, \ldots, f_m \), then the final CHR rule for \( f \) is

\[
f(x) \iff f_1(x), \ldots, f_m(x)
\]

Note the lack of justifications, which will be collected from the rules for \( f_1, \ldots, f_m \).

5.3 Recursive Functions

If we were to naively apply the scheme outlined before the translation of recursive programs, our type inference procedure would become undecidable. In short, a CHR derivation in such a situation would never terminate.

**Example 15. Consider the program:**

\[
\begin{align*}
f \quad (x, y) &= \ast \quad g \quad (x, y) = \ast \quad \text{True} \\
& \quad \text{g} \quad (x, y) = \ast \quad \text{True}
\end{align*}
\]

Applying the standard translation, we would generate something like the following.

\[
\begin{align*}
f(t_f) & \iff g(t_y), t_g = (t_x, t_y) \rightarrow \text{t_r}, t_f = (t_x, t_y) \rightarrow \text{t_r} \\
g(t_y) & \iff f(t_y), t_f = (t_y, t_x) \rightarrow \text{t_r}, t_g = (t_x, t_y) \rightarrow \text{t_r}
\end{align*}
\]
with the knowledge that the calls to a CHR derivation involving either rule, it is clear that it would never terminate.

To circumvent this problem, we enforce monomorphic recursion for functions without type annotations in Chameleon. Consequently, this allows us to replace the user constraints involved in cycles with monomorphic types. We do this by unfolding cyclical user constraints.

**Example 16.** We return to the two CHRs generated above, with the knowledge that the calls to f and g within their bodies are involved in a cycle.

We begin with the rule for f above, and apply the simplification rule for g to the rhs constraints, obtaining the following:

\[
\begin{align*}
    f(t_f) & \iff f(t'_f), t'_f = (t'_x, t'_y) \rightarrow t'_v, t'_x = (t'_x, t'_y) \rightarrow t'_v, \\
    t'_v & = t'_w, t'_w = (t_x, t_y) \rightarrow t_r, t_f = (t_x, t_y) \rightarrow t_r,
\end{align*}
\]

The newly added constraints are shown in boldface. The constraint \(t'_v = t'_w\) represents the matching of the type in the g user constraint and the type in the head of the g rule.

Finally, to break the cycle we replace the call to \(f\) with a constraint asserting that \(t'_f\) is the same type as we have already found for \(f\) - hence the monomorphism.

\[
f(t_f) \iff t'_f = t_r, t'_y = (t'_y, t'_x) \rightarrow t'_v, t'_y = (t'_x, t'_y) \rightarrow t'_r,
\]

The same procedure would then be carried out for the g rule.

**Example 17.** Consider the function:

\[
\begin{align*}
f & : [a] \rightarrow \text{Bool} \\
f \text{[]} & = \text{True} \\
f (x:xs) & = f \ [xs]
\end{align*}
\]

Note that the type annotation is necessary. In such a case, our translation (simplified) yields the following.

\[
\begin{align*}
f_a(t) & \iff t = [a] \rightarrow \text{Bool} \\
f(t) & \iff f_1(t), f_2(t) \\
f_1(t) & \iff t = [a] \rightarrow \text{Bool} \\
f_2(t) & \iff t = [a] \rightarrow t_1, t_2 = [a] \rightarrow t_1, f_a(t_2)
\end{align*}
\]
5.4 Overloading

For an in-depth treatment of the translation of Haskell-class and instance declarations to CHRs we refer the interested reader to [24, 8].

The translation from declarations to CHRs is

\[
\text{instance } (C \Rightarrow TC) \sigma_i, \text{ where } \quad TC \ t \leftrightarrow C_0 \\
\text{class } (C \Rightarrow TC \ x) \sigma_i, \text{ where } \quad TC \ x \leftrightarrow C_i \\
\text{f :: } (C \Rightarrow T) \sigma_i, \text{ where } \quad f_a(y) \leftrightarrow g = \ t, C_{12}, (TC \ x) \sigma_i
\]

The appropriate location \((l_0, l_1 \text{ or } l_2)\) is added as justification to all constraints on the right hand side. A missing constraint \(C\) is treated as \(True\). Note that we use upper-case letters for user-defined type class constraints, and lower-case letters for user-defined constraints referring to function definitions.

**Example 18.** Given the class and instance declarations below,

\[
\text{class } (\text{Eq } a)_{50} \text{ where } \\
(\Rightarrow) :: a \rightarrow a \rightarrow \text{Bool}_{51} \\
\text{class } (\text{Ord } a)_{52} \text{ where } \\
(>) :: a \rightarrow a \rightarrow \text{Bool}_{53} \\
\text{instance } (\text{Ord } a)_{54} \text{ where } \\
(\text{Eq } a)_{50} \\
\text{Ord } a_{52} \\
\text{Ord } [a]_{54} \\
\text{Ord } \text{Bool}_{55} \text{ where } \\
\text{True } \rightarrow \text{False } \leftrightarrow \text{True} \\
\text{False } \rightarrow \text{False } \leftrightarrow \text{False}
\]

we generate the following CHRs

\[
\text{Eq } a \quad \Rightarrow \quad \text{True} \\
\text{eq}_a(t_51) \quad \leftrightarrow \quad (t_51 = a \rightarrow a \rightarrow \text{Bool}_{51}, (\text{Eq } a)_{51} \\
\text{Ord } a \quad \Rightarrow \quad (\text{Eq } a)_{51} \\
gt_a(t_53) \quad \leftrightarrow \quad (t_53 = a \rightarrow a \rightarrow \text{Bool}_{53}, (\text{Ord } a)_{53} \\
\text{Ord } [a] \quad \Rightarrow \quad (\text{Ord } a)_{54} \\
\text{Ord } \text{Bool} \quad \Rightarrow \quad \text{True}
\]

We assume \(eq\) represents the type of Eq’s member function \((\Rightarrow)\) and \(gt\) represents the type of Ord’s member function \((>)\). Note that CHRs arising from the type annotations appearing in the classes have a missing implicit class constraint added.

Note we would also generate constraints for the code defining the instance methods for \(\Rightarrow\) and check this versus the annotation constraints for \(gt\). [□]

The proof of correctness of these rules in modeling the class constraints can be found in [8]. Note that our type debugging approach also immediately extends to more complicated approaches to overloading that can be expressed as CHRs [24].

We generally assume that CHRs are confluent. A set of CHRs is confluent if any sequence of derivation steps leads to the same final constraint store. This condition holds trivially for CHRs generated from the Hindley/Milner subset of our language. The same is true for any valid set of Haskell 98 [10] class and instance declarations.

6. TYPE INFERENCE VIA CHR SOLVING

Consider type inference for a function definition \(f = e\). We execute the goal \(f(t, l)\) using the CHR program \(P\) created, i.e. \(f(t, l) \rightarrow^* C\), and build \(\phi\), the most general unifier of \(C\). Let \(a = \{\phi(C_u) \setminus \{\phi(\ell_l)\}\}\). These are the variables we will quantify over; we specifically exclude types of \(\lambda\)-bound variables. We can then build the type scheme, \(\phi : \forall \phi C_u \Rightarrow \phi\).

Note that in our scheme we are a bit more “lazy” in detecting type errors compared to other formulations.

**Example 19.** Consider

\[
e = \text{let } f = \text{True True in False}
\]

Our (simplified) translation to CHRs yields

\[
f(t) \quad \leftrightarrow \quad t_1 = \text{Bool}, t_1 = t_2 \rightarrow t_3, t_2 = \text{Bool}, t_3 = t \\
e(t) \quad \leftrightarrow \quad t = \text{Bool}
\]

For simplicity, we omit justifications and the \(l\) parameter. Note that type inference for expression \(e\) succeeds, although function \(f\) is ill-typed. There is no occurrence of \(\text{True} \ in \ the \ body\), hence we never execute the CHR belonging to \(f\).

In a traditional approach, type inference for \(e\) proceeds by first inferring the type of \(f\) immediately detecting that \(f\) is not well-typed. Note that our approach is still type-safe for a lazy language. Additionally, we could require that all defined functions must be type correct, by simply executing the corresponding CHRs. [□]

**Example 20.** Consider the following program, together with the class and instance declarations of Example 18

\[
lteq_{10} x_1 y_2 = (\text{not} \_7 ((x_4 \geq 5 \ y_6)_{10})) \\
\text{not} :: \text{Bool} \rightarrow \text{Bool}_{16}
\]

where \((\geq)\) is part of the \text{Ord} class. The translation process yields (again for simplicity we ignore the \(\lambda\)-bound variables argument):

\[
lteq(t_{10}) \quad \leftrightarrow \quad (t_1 = t_4)_{1}, (t_2 = t_5)_{2}, g(t_3)_{3}, (t_4 = t_3)_{4}, (t_3 = t_1 \rightarrow t_5)_{5}, (t_1 = t_6)_{6}, \\text{not}(t)_{7}, (t_5 = t_1 \rightarrow t_6)_{8}, (t_7 = t_8 \rightarrow t_9)_{9}, (t_{10} = t_1 \rightarrow t_2 \rightarrow t_3)_{10} \\
\text{not}(t_{16}) \quad \leftrightarrow \quad (t_{16} = \text{Bool} \rightarrow \text{Bool}_{16})
\]

These rules are generated directly from the program text.

Type inference for \((\text{lteq}_{22} \ [w_{7} t_{18}]_{23})\) generates the constraints

\[(t_{17} = t_{17}, (t_{18} = (t_{17}))_{18}, \text{lteq}(t_{22})_{22}, (t_{22} = t_{18} \rightarrow t_{23})_{23}\)

This is the initial constraint which we run the CHR program on. For the first step, we find \(E = \emptyset\) and \(D = \{\text{lteq}(t_{22})\}\) means we can apply the first rule above leading to

\[(t_{17} = t_{17}, (t_{18} = (t_{17}))_{18}, t_{10} = t_{22}, (t_1 = t_3)_{1,22}, (t_2 = t_5)_{2,22}, g(t_3)_{3,22}, (t_4 = t_3)_{4,22}, (t_3 = t_4 \rightarrow t_5)_{5,22}, (t_6 = t_9)_{6,22}, \text{not}(t)_{7,22}, (t_7 = t_8 \rightarrow t_9)_{8,22}, (t_9 = t_9)_{9,22}, (t_{10} = t_1 \rightarrow t_2 \rightarrow t_3)_{10,22}, (t_{22} = t_{18} \rightarrow t_{23})_{23}\)

For brevity, we show the whole derivation in a simplified form, just showing \(\theta(C_u) \land \ell_l = \theta(t_{23})\) where \(\theta\) is the mgu of \(C_u\) for \(C\) at each step, and omit justifications. That is, we only show the user-defined constraints and the top type
variable $t_{23}$, under the effect of the equations in $C$ ignoring justifications.

\[
\begin{align*}
\text{letq}(\lfloor t_w \rfloor \rightarrow t_{23}), t_{23} &= t_{23} \\
\text{not} t_{23} &\rightarrow t_{18} \\
\text{gt}((t_{18} \rightarrow t_{9})), t_{23} &= t_{23} \\
\text{Ord} \{t_{18} \rightarrow t_{2} \rightarrow \text{Bool}\}, t_{23} &= t_{23} \rightarrow \text{Bool} \\
\text{Ord} a, t_{23} &= \lfloor t_w \rfloor \rightarrow \text{Bool} \\
\text{Ord} a, t_{23} &\rightarrow \text{Eq} t_w, t_{23} &= \lfloor t_w \rfloor \rightarrow \text{Bool} \\
\end{align*}
\]

In other words the type inferred for the original expression
\[
(\text{Ord } a, \text{ Eq } a) \Rightarrow \lfloor a \rfloor \rightarrow \text{Bool}
\]
Note that we are more “verbose” than e.g. Hugs [15] which would report $\text{Ord } a \Rightarrow \lfloor a \rfloor \rightarrow \text{Bool}$. Clearly, the constraint $\text{Eq } a$ is “redundant”, since every instance of $\text{Ord } a$ must be an instance of $\text{Eq}$ as specified by the class declaration for $\text{Ord}$. In [24], we show how to remove such redundant constraints. However, for type debugging purposes it is desirable to keep all constraints for better type explanations.

The fourth step in the derivation is the only one requiring a non-empty set $E$ of equations to justify the match. The constraint $D = \{ (\text{Ord } a_1)_{(3,22,53)}, (\text{Ord } a_2)_{(3,22,53)} \}$ matches the left hand side of the rule $\text{Ord} \{a_2\} \Leftarrow (\text{Ord } a_2)_{(3,22,53)}$. The minimal set of equations $E \subseteq C$ where $\theta = \text{mgu}(E)$ is such that $\theta(a_1)$ has the form $[t']$ is

\[
\begin{align*}
(t_1 = t_2)_{(1,22)}, (t_{23} = a_1 \rightarrow a_1 \rightarrow \text{Bool})_{(3,22,53)}, t_3 = t_{53}, \\
(t_4 = t_5)_{(4,22)}, (t_4 = t_2 \rightarrow t_5)_{(5,22)}, t_1 = t_{22}, t_1 = [t_{17}]_{18}, \\
(t_{10} = t_1 \rightarrow t_2 \rightarrow t_9)_{(10,22)}, (t_{22} \rightarrow t_2 \rightarrow t_{23})_{23}
\end{align*}
\]

The total justifications of $E \cup D$ are $\{1, 3, 4, 5, 10, 18, 22, 23, 53\}$. Hence we replace the constraint $\text{Ord } a_1_{(3,22,53)}$ by

\[
[a_2] = a_1, (\text{Ord } a_2)_{(3,22,53)}
\]

We can state soundness and completeness of type inference for the system described in Figure 1. We assume that the type of predefined functions is recorded as a CHR. For example, $\text{map} : (a \rightarrow b) \rightarrow [a] \rightarrow [b]$ is represented by

\[
\text{map}(t, l) \iff t = (a \rightarrow b) \rightarrow ([a] \rightarrow [b]).
\]

**Theorem 1 (Soundness and Completeness).** Let $P_i$ be a set of CHRs describing all predefined functions, $\Gamma$ be an environment containing all free variables, $e$ be an expression and $t$ be a type. Then, we have that $\Gamma \vdash e \rightarrow t \iff \Gamma' \vdash e \rightarrow \text{Cons}(C \mid t')$ for some constraint $C$ and type $t'$ if $\Gamma' \vdash \text{Def}(P_i)$ for some set $P_i$ of CHRs such that $C \rightarrow t_1 \cup P_i$ and $\phi \rightarrow t$ with $\phi$ m.g.u. of $D$ where we consider $\text{free}(\Gamma)$ as Skolem constants.

The theorem can be reestablished for each of the type extensions considered.

Note that in case of type annotations, type inference needs to perform a subsumption check. We compare a type annotation for function $f$ versus the function definition by simply testing the following. We execute $f_a(t,l) \rightarrow t_1$ and $f(t,l) \rightarrow t_2$ and check that $t_1 \subseteq C_1, C_2$. The correctness of this check is proved in [24].

In case of overloading, we must also ensure that type schemes are unambiguous. For Haskell 98 this equates to checking that variables appearing within the context of the type scheme must also appear within the type. Not enforcing this condition would make the semantics of such programs non-deterministic. For details, we refer to [24].

## 7. Constraint Operations

The type debugger make use of two essential manipulations of the constraints generated from the CHR derivation: finding a minimal unsatisfiable subset of unsatisfiable constraint set, and finding a minimal subset that implies some give constraint (which may be used if the constraints are satisfiable or unsatisfiable). Justifications attached to those minimal sets refer to problematic program locations.

### 7.1 Minimal Unsatisfiable Subsets

Assume type inference fails. That is, we have that $C \rightarrow^*_p D$ for some constraint $C$ and $D$ where $D$ is unsatisfiable. For $D$ to be unsatisfiable it must be that $D_1$ is unsatisfiable, since user-defined constraints only contribute new equations.

We are interested in finding a minimal subset $E$ of $D_1$ such that $E$ is unsatisfiable. An unsatisfiable set is minimal if the removal of any constraint from that set leaves it satisfiable. The Chameleon system simply finds an arbitrary minimal unsatisfiable subset. We also determine which constraints in this set are present in all minimal unsatisfiable subsets.

We can naively determine minimal unsatisfiable subsets by testing each possible subset. This is impractical.

Using an incremental resolution solver (as all unification algorithms are) we can quickly determine a minimal unsatisfiable subset of $D$ by adding the equations one at a time and detecting the first time the set is unsatisfiable. The last added equation must be involved in the minimal unsatisfiable subset.

Applying this principle repeatedly results in:

\[
\text{min_unsat}(D) \Leftarrow \emptyset
\]

\[
\text{while satisfiable}(M) \{ \\
\quad C := M \\
\quad \text{while satisfiable}(C) \{ \\
\quad \quad \text{let } e \in D - C; C := C \cup \{e\} \\
\quad D := C; M := M \cup \{e\} \\
\quad \}\ \text{return } M
\]

We can straightforwardly determine which constraints $e \in M$ must occur in all minimal unsatisfiable subsets, since this is exactly those where $D - \{e\}$ is satisfiable. The complexity (for both checks) is $O(|D|^2)$ using an incremental unification algorithm. A detailed analysis of the problem of finding all minimal unsatisfiable constraints can be found in [7].

Ultimately, we are interested in the justifications attached to minimal unsatisfiable constraints. This will allow us to identify problematic locations in the program text.

**Example 21.** Consider the final constraint of Example 1.

\[
(t_1 = \text{Char})_{(1,8)}, t_2 = t_7, (t_5 = \text{Bool})_{(5,2,8)}, (t_6 = \text{Bool})_{(6,2,8)}, \\
(t_7 = t_5 \rightarrow t_6)_{(7,2,8)}, (t_2 = t_1 \rightarrow t_3)_{(8,3)}, (t_4 = t_3)_{(4,8)}
\]

The system of constraints is detected as unsatisfiable as the second last constraint $(t_2 = t_1 \rightarrow t_3)_{(3,8)}$ is added. Hence $(t_4 = t_3)_{(4,8)}$ can be excluded from consideration. Solving from the beginning, starting with $(t_2 = t_1 \rightarrow t_3)_{(3,8)}$, unsatisfiability is detected at $(t_2 = t_5 \rightarrow t_6)_{(7,2,8)}$. In the next iteration, starting with $(t_7 = t_5 \rightarrow t_6)_{(7,2,8)}$ and $(t_2 = t_1 \rightarrow t_3)_{(3,8)}$, unsatisfiability is detected at $(t_6 = \text{Bool})_{(5,2,8)}$. Therefore, $(t_6 = \text{Bool})_{(6,2,8)}$ can be excluded.

The final result $M$ is

\[
(t_1 = \text{Char})_{(1,8)}, t_2 = t_7, (t_5 = \text{Bool})_{(5,2,8)}, \\
(t_7 = t_5 \rightarrow t_6)_{(7,2,8)}, (t_2 = t_1 \rightarrow t_3)_{(3,8)}
\]

Note that $M$ is the only minimal unsatisfiable constraint for this example.
7.2 Minimal Implicants

We are also interested in finding minimal systems of constraints that ensure that a type has a certain shape. Assume that $C \rightarrow_{\gamma} D$ where $\models D \supseteq \exists a. F$, unexpectedly, where $F$ is a conjunction of equations. We want to identify a minimal subset $E$ of $D$ such that $\models E \supseteq \exists a. E$. The algorithm for finding minimal implicants is highly related to that for minimal unsatisfiable subsets.

The code for min_impl is identical to min_unsat except the test satisfiable$(S)$ is replaced by $\neg$implies$(S, \exists a. D')$.

\[
\begin{align*}
\text{min_impl}(D) \\
&\text{while } \neg \text{implies}(M, \exists a. D') \\
&\quad \text{C := M} \\
&\quad \text{while } \neg \text{implies}(C, \exists a. D') \\
&\quad \quad \{ \text{let } e \in D - C; C := C \cup \{e\} \} \\
&\quad D := C; M := M \cup \{e\} \\
&\text{return } M
\end{align*}
\]

The test implies$(M, \exists a. D')$ can be performed as follows. If $D'$ is a system of equations only, we simply add them and check that no variable apart from those in $\exists a$ is bound.

If $D'$ includes user defined constraints, then for each user-defined constraint $c_i \in D'$ we nondeterministically choose a user-defined constraint $c'_i \in M$. We then check that implies$(M, \exists a. (D' \cup \{c_i = c'_i\})$ holds as above. We need to check all possible choices for $c'_i$ (although we can omit those which obviously lead to failure, e.g. $c_i = Eq a$ and $c'_i = Ord b$).

8. RELATED WORK

The most conservative approach to improving type error information involves modifying the order in which substitutions take place within traditional inference algorithms. The standard algorithm, $\mathcal{W}$, tends to find errors too late in its traversal of a program [17, 29], since it delays substitutions until as late as possible. $\mathcal{W}$ has been generalized [17] so that the point at which substitutions are applied can be varied. Despite this, there are cases where it is not clear which variant provides the most appropriate error report. Moreover, all of these algorithms suffer from a left-to-right bias when discovering errors during abstract syntax tree (AST) traversal.

One way to overcome this problem, as we have seen, is to avoid the standard inference algorithms altogether and focus directly on the constraints involved. Although our work bears a strong resemblance to [11, 12, 13], our aims are different. We attempt to explain errors involving advanced type system features, such as overloading, whereas [13], who are developing a beginner-friendly version of Haskell, choose to ignore such features by design. Furthermore, they focus on producing non-interactive error messages, and do not consider mechanisms for providing type explanations.

In [19], graphs are used to represent type information, again, independently of any particular program traversal. This work allows generation of potentially more useful type error messages, again without any opportunity for user interaction.

A number of "error explanation systems" [2, 5, 28] allow the user to examine the process by which specific types are inferred. By essentially recording the effects of the inference procedure on specific types a step at a time, a complete history can be built up. One common shortcoming of such systems is the excessive size of explanations. Although complete, such explanations are full of repetitive and redundant information which can be a burden to deal with. Furthermore, since these systems are layered on top of an existing inference algorithm, they suffer from the same AST traversal bias. In contrast, when asked to explain why an expression has a particular type, our system finds precisely those locations which have contributed.

Chitil [3] describes a compositional type explanation system based on the idea of principal typings. In his system a user can explore the types of subexpressions by manually navigating through the inference tree. This is very similar to our form of declarative debugging (Section 2.6). Note that our form of type explanation allows us to automatically identify contributing program locations.

Independently, Haack and Wells [9] also discuss finding of minimal unsatisfiable subsets which allows them to find problematic program locations. However, they only consider error explanations. That is, in their system it is not possible to explain why functions have a type of a certain shape. Furthermore, their approach applies to the Hindley/Milner system only whereas our approach is applicable to Haskell-style type classes and its various extensions.

9. CONCLUSION

We have presented a flexible type debugging scheme for Hindley/Milner typable programs which also includes Haskell-style overloading. The central idea of our approach is to translate the typing problem to a constraint problem, i.e. a set of CHRs. Type inference is phrased in terms of CHR solving. Our approach has the advantage that we are not dependent on a fixed traversal of the abstract syntax tree. Constraints can be processed in arbitrary order which makes a flexible traversal of the syntax tree possible.

In case of a type error (or unexpected result), we find minimal unsatisfiable constraints (minimal implications). Justifications, i.e. program locations, attached to constraints allow us to identify problematic program expressions. The approach has been fully implemented [26] and can be used as a front-end to any existing Haskell system.

There is much further work to do in improving the system. This includes adding features such as: allowing the user to trace the CHR type inference derivation, and explaining each step in the derivation, and using the minimal unsatisfiable subsets to generate better error messages. In particular, we plan to include some heuristics to catch common errors. The Helium [13] programming environment includes a database of common mistakes which is searched for a match when a type error occurs. This allows meaningful error messages and suggestions on how to fix the error to be presented. Using minimal unsatisfiable subsets to search in the database should allow us to detect more generic common mistakes.

10. REFERENCES


