Where is ML type inference headed? Constraint solving meets local shape inference

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Types are good

A type is a concise description of the behavior of a program fragment.

Typechecking provides safety or security guarantees.

It also encourages modularity and abstraction.

Type inference is good

Types can be extremely cumbersome when they have to be explicitly and repeatedly provided.

This leads to (partial or full) type inference...

... which is sometimes hard, but so... addictive.

Constraints are elegant

Type inference problems are naturally expressed in terms of constraints made up of predicates on types, conjunction, existential and universal quantification, and possibly more.

This allows reducing type inference to constraint solving.

Mandatory type annotations can help

Constraint solving can be intractable or undecidable for some (interesting) type systems.

In that case, mandatory type annotations can help. Full type inference is abandoned. In return, the reduction of (now partial) type inference to constraint solving is preserved.

One might wish to go further...

Stratified type inference

Local shape inference can be used to propagate type information in ad hoc ways through the program and automatically produce some of the required annotations.

This leads to stratified type inference, a pragmatic approach to hard type inference problems.

Overview

The talk is planned as follows:

- 1. Constraint-based type inference for ML
- 2. Stratified type inference for generalized algebraic data types

Part I

Type inference for ML

The simply-typed λ -calculus

Hindley and Milner's type system

Specification

The simply-typed λ -calculus is specified using a set of rules that allow deriving judgements:

Var
$$\Gamma \vdash x : \Gamma(x)$$
Abs
$$\frac{\Gamma; x : \tau_1 \vdash e : \tau_2}{\Gamma \vdash \lambda x.e : \tau_1 \to \tau_2}$$

$$\frac{\Gamma; x : \tau_1 \vdash e : \tau_2}{\Gamma \vdash e_1 : \tau_1 \to \tau_2}$$

$$\frac{\Gamma \vdash e_1 : \tau_1 \to \tau_2}{\Gamma \vdash e_1 : e_2 : \tau_2}$$

The specification is syntax-directed.

Constraints

In order to reduce type inference to constraint solving, we introduce a constraint language:

$$C ::= \tau = \tau \mid C \wedge C \mid \exists a.C$$

Constraints are interpreted by defining when a valuation ϕ satisfies a constraint C.

Constraint solving is first-order unification.

Constraint generation

Type inference is reduced to constraint solving by defining a mapping $[\cdot]$ of candidate judgements to constraints.

```
\begin{bmatrix}
\Gamma \vdash x : \tau
\end{bmatrix} = \Gamma(x) = \tau \\
\begin{bmatrix}
\Gamma \vdash \lambda x.e : \tau
\end{bmatrix} = \exists a_1 a_2.(\begin{bmatrix}
\Gamma; x : a_1 \vdash e : a_2
\end{bmatrix} \land a_1 \rightarrow a_2 = \tau) \\
\begin{bmatrix}
\Gamma \vdash e_1 e_2 : \tau
\end{bmatrix} = \exists a.(\begin{bmatrix}
\Gamma \vdash e_1 : a \rightarrow \tau
\end{bmatrix} \land \begin{bmatrix}
\Gamma \vdash e_2 : a
\end{bmatrix})
```

Constraints, revisited

How about letting the constraint solver, instead of the constraint generator, deal with *environments?*

Let's enrich the syntax of constraints:

$$C ::= \dots \mid x = \tau \mid def x : \tau in C$$

The idea is to interpret constraints in such a way as to validate the equivalence law

$$def x : \tau in C \equiv [\tau/x]C$$

The def form is an explicit substitution form.

Constraint generation, revisited

Constraint generation is now a mapping of an expression e and a type τ to a constraint $[e:\tau]$.

$$\begin{bmatrix} x : \tau \end{bmatrix} = x = \tau
\begin{bmatrix} \lambda x.e : \tau \end{bmatrix} = \exists a_1 a_2. \begin{pmatrix} def x : a_1 \text{ in } \llbracket e : a_2 \rrbracket \\ a_1 \to a_2 = \tau \end{pmatrix}
\begin{bmatrix} e_1 e_2 : \tau \rrbracket = \exists a. (\llbracket e_1 : a \to \tau \rrbracket \land \llbracket e_2 : a \rrbracket)$$

Look ma, no environments!

The point of introducing the def form will become apparent in Hindley and Milner's type system...

The simply-typed λ -calculus

Hindley and Milner's type system

Specification

Three new typing rules are introduced in addition to those of the simply-typed λ -calculus:

Gen
$$\frac{\Gamma \vdash e : \tau \quad \bar{a} \# ftv(\Gamma)}{\Gamma \vdash e : \forall \bar{a}.\tau} \qquad \frac{\Gamma \vdash e : \forall \bar{a}.\tau}{\Gamma \vdash e : [\vec{t}/\vec{a}]\tau}$$
Let
$$\frac{\Gamma \vdash e_1 : \sigma \quad \Gamma; x : \sigma \vdash e_2 : \tau}{\Gamma \vdash let \ x = e_1 \text{ in } e_2 : \tau}$$

Type schemes now occur in environments and judgements.

Constraints

Let's extend the syntax of constraints so that a variable x can stand for a *type scheme*.

To avoid mingling constraint generation and constraint solving, we allow type schemes to carry constraints.

Turning a constraint into a (constrained) type scheme is then a purely syntactic construction—no solving is required.

Constraints, continued

The syntax of constraints and constrained type schemes is:

$$C ::= \tau = \tau \mid C \land C \mid \exists a.C \mid x \leq \tau \mid def \ x : \varsigma \text{ in } C$$

$$\varsigma ::= \forall \overline{a}[C].\tau$$

The idea is to interpret constraints in such a way as to validate the equivalence laws

$$def x : \varsigma in C \equiv [\varsigma/x]C$$
$$(\forall \bar{a}[C].\tau) \preceq \tau' \equiv \exists \bar{a}.(C \land \tau = \tau')$$

Constraint generation

Constraint generation is modified as follows:

$$[\![x:\tau]\!] = x \leq \tau$$

$$[\![let x = e_1 \text{ in } e_2:\tau]\!] = def x : \forall a [\![\![e_1:a]\!]].a \text{ in } [\![e_2:\tau]\!]$$

The constrained type scheme $\forall a[[e_1:a]].a$ is principal for $e_1...$

Statement

Theorem (Soundness and completeness)

Let Γ be an environment whose domain is fv(e). The expression e is well-typed relative to Γ iff

$$def \Gamma in \exists a. [e:a]$$

is satisfiable.

Taking constraints seriously

Constraints are suitable for use in an efficient and modular implementation, because:

- constraint generation has linear complexity;
- constraint generation and constraint solving are separate;
- the constraint language remains simple as the programming language grows.

Part II

Generalized algebraic data types

Introducing generalized algebraic data types

Typechecking: MLGI

Simple, constraint-based type inference: MLGX

Local shape inference

Example

Here is typed abstract syntax for a simple object language.

```
Lit :: int \rightarrow term int

Inc :: term int \rightarrow term int

IsZ :: term int \rightarrow term bool

If :: \forall a.term bool \rightarrow term a \rightarrow Snd :: \forall a\beta.term a \rightarrow term a \rightarrow Snd :: \forall a\beta.term a \rightarrow term a \rightarrow Snd :: \forall a\beta.term a \rightarrow term a \rightarrow Snd :: \forall a\beta.term a \rightarrow term a \rightarrow Snd :: \forall a\beta.term a \rightarrow term a \rightarrow Snd :: \forall a\beta.term a \rightarrow term a \rightarrow term a \rightarrow Snd :: \forall a\beta.term a \rightarrow term a \rightarrow term a \rightarrow term a \rightarrow Snd :: \forall a\beta.term a \rightarrow term a \rightarrow term
```

This is not an ordinary algebraic data type...

Example, continued

This definition allows writing an evaluator that performs no tagging or untagging of object-level values, that is, no runtime checks:

From type inference to constraint solving

In the presence of generalized algebraic data types, reducing type inference to constraint solving remains reasonably straightforward.

For eval, the constraint looks like this, after several simplification steps:

$$\forall a. \left(\begin{array}{l} a = \text{int} \Rightarrow \text{int} = a \text{ // Lit} \\ \dots \\ \forall a_1 a_2 . a = a_1 \times a_2 \Rightarrow a_1 \times a_2 = a \text{ // Pair} \\ \dots \end{array} \right)$$

This eventually simplifies down to true, so eval is well-typed.

It looks as if there is no problem?

Implications of implication

Adding implication to the constraint language yields the first-order theory of equality of trees, whose satisfiability problem is decidable, but intractable.

For eval, solving seemed easy because enough explicit information was available.

Furthermore, introducing implication means that constraints no longer have most general unifiers, as the next example shows...

Implications of implication, continued

What types does this function admit?

Eq::
$$\forall a.eq \ a \ a$$

cast =
 $\forall a\beta.\lambda(w:eq \ a \ \beta).\lambda(x:a).$
case w of
 $Eq \rightarrow (*a = \beta *) x$

Implications of implication, continued

All three type schemes below are correct:

$$\forall a\beta.eq\ a\ \beta \rightarrow a \rightarrow a$$

 $\forall a\beta.eq\ a\ \beta \rightarrow a \rightarrow \beta$
 $\forall \gamma.eq\ int\ bool\ \rightarrow int\ \rightarrow \gamma$

but *none* is principal! The principal *constrained* type scheme produced by constraint solving would be

$$\forall a\beta y [a = \beta \Rightarrow a = y].eq \ a \ \beta \rightarrow a \rightarrow y$$

which indeed subsumes the previous three.

The system does not have principal types in the standard sense.

A solution

I am now about to present a solution where principal types are recovered by means of mandatory type annotations and where a local shape inference layer is added so as to allow omitting some of these annotations.

This is joint work with Yann Régis-Gianas.

Introducing generalized algebraic data types

Typechecking: MLGI

Simple, constraint-based type inference: MLGX

Local shape inference

MLGI

Let's first define the programs that we deem sound and would like to accept, without thinking about type inference.

This is MLGI—ML with generalized algebraic data types in implicit style.

MLGI is Core ML with polymorphic recursion, generalized algebraic data types, and explicit type annotations.

Specification

MLGI's typing judgments take the form

 $E, \Gamma \vdash e : \sigma$

where E is a system of type equations.

Most of the rules are standard, modulo introduction of E...

Specification, continued

E is exploited via implicit type conversions:

$$\frac{a = \text{int}, \Gamma \vdash i : \text{int} \qquad a = \text{int} \Vdash \text{int} = a}{a = \text{int}, \Gamma \vdash i : a}$$

The symbol I stands for constraint entailment.

Specification, continued

```
Pair a b: term a \vdash (a_1 a_2, a = a_1 \times a_2, a : a_1; b : a_2)
\underbrace{a_1 a_2 \# \text{ftv}(\Gamma, a)}_{\text{true}, \Gamma} = \underbrace{a_1 \times a_2, (\Gamma; a : a_1; b : a_2) \vdash e : a}_{\text{true}, \Gamma}
```

Inside each clause, confronting the pattern with the (actual) type of the scrutinee yields new (abstract) type variables, type equations, and environment entries.

Determining E and inferring types are interdependent activities...

Introducing generalized algebraic data types

Typechecking: MLGI

Simple, constraint-based type inference: MLGX

Local shape inference

MLGX

Let's require sufficiently many type annotations to ensure that E is known at all times, without any guessing. Let's also make all type conversions explicit.

This is MLGX—ML with generalized algebraic data types in explicit style.

Specification

$$E, \Gamma \vdash (e : \text{term } a) : \text{term } a$$

$$\forall i \quad E, \Gamma \vdash (p_i : \text{term } a).e_i : \text{term } a \rightarrow a$$

$$E, \Gamma \vdash \text{case } (e : \text{term } a) \text{ of } p_1.e_1 \dots p_n.e_n : a$$

We require a type annotation at case constructs and pass it down to the rule that examines individual clauses...

Specification, continued

The rule that checks clauses now exploits the type annotation:

Pair a b: term
$$a \vdash (a_1 a_2, a = a_1 \times a_2, a : a_1; b : a_2)$$

$$a_1 a_2 \# ftv(\Gamma, a) \qquad a = a_1 \times a_2, (\Gamma; a : a_1; b : a_2) \vdash e : a$$

$$true, \Gamma \vdash (Pair \ a \ b : term \ a) \cdot e : term \ a \to a$$

The pattern is now confronted with the type annotation to determine which new type equations arise. No guessing is involved.

Specification, continued

E is now exploited only through an explicit coercion form:

$$\frac{a = \text{int}, \Gamma \vdash i : \text{int} \qquad a = \text{int} \Vdash \text{int} = a}{a = \text{int}, \Gamma \vdash (i : (\text{int} \triangleright a)) : a}$$

This rule is syntax-directed.

Type inference for MLGX

Type inference for MLGX decomposes into two separate tasks:

- compute E everywhere and check that every explicit coercion is valid;
- ▶ forget E and follow the standard reduction to constraint solving. A coercion (int \triangleright a) is just a constant of type int \rightarrow a.

No implication constraints are involved. MLGX has principal types.

In short, MLGX marries type inference for Hindley and Milner's type system with typechecking for generalized algebraic data types. I believe its design is robust.

Programming in MLGX

In MLGX, eval is written:

```
\mu(eval: \forall a. term \ a \rightarrow a). \forall a. \lambda t.
case (t: term \ a) \text{ of}
| \text{ Lit } i \rightarrow (i: (int \triangleright a))
| \text{ Inc } t \rightarrow (eval \ t+1: (int \triangleright a))
| \text{ Is} Z \ t \rightarrow (eval \ t=0: (bool \triangleright a))
| \text{ If } b \ t \ e \rightarrow \text{ if } eval \ b \text{ then } eval \ t \text{ else } eval \ e
| \text{ Pair } a_1 \ a_2 \ a \ b \rightarrow ((eval \ a, eval \ b): (a_1 \times a_2 \triangleright a))
| \text{ Fst } \beta_2 \ t \rightarrow \text{ fst } (eval \ t)
| \text{ Snd } \beta_1 \ t \rightarrow \text{ snd } (eval \ t)
```

This is nice, but redundant... how about some local shape inference?

Introducing generalized algebraic data types

Typechecking: MLGI

Simple, constraint-based type inference: MLGX

Local shape inference

Shapes

Shapes are defined by

$$s := \bar{\gamma}.\tau$$

The flexible type variables $\bar{\gamma}$ (bound within τ) represent unknown or polymorphic types.

That is, the shape $\gamma.\gamma \to \gamma$ adequately describes the integer successor function as well as the polymorphic identity function.

This shape is much more precise than $\gamma_1\gamma_2.\gamma_1 \rightarrow \gamma_2$, which describes any function.

Shapes, continued

Shapes can have *free* type variables; these are interpreted as *known* types. For instance, the shape

$$\gamma.a \times \gamma$$

describes a pair whose first component has type a, where the type variable a was explicitly and universally bound by the programmer, and whose second component has unknown type.

Ordering shapes

Shapes are equipped with a standard instantiation ordering.

For instance,

$$(\gamma_1.a \times \gamma_1) \leq (\gamma_2.a \times (a \rightarrow \gamma_2))$$

The uninformative shape γ, γ , written \bot , is the least element.

Ordering shapes, continued

When two shapes have an upper bound, they have a least upper bound, computed via first-order unification.

For instance,

$$(\gamma.\gamma \rightarrow \gamma) \sqcup (\gamma.int \rightarrow \gamma) = int \rightarrow int$$

This allows local shape inference to find that "applying the identity function to an integer yields an integer" — reasoning that requires instantiation.

Yet, this use of unification is *local*, because flexible type variables are never shared between shapes.

Algorithm Z, judgements

Here is a very rough overview of a shape inference algorithm.

Judgements take the form

$$E, \Gamma \vdash e \downarrow s \uparrow s' \leadsto e'$$

where Γ (which maps variables to shapes) and s are provided, while s' is inferred and at least as informative, that is, $s \leq s'$ holds.

Algorithm Z, mission statement

The transformed term e' is identical to e, except

- ▶ type coercions are inserted at variables and at case clauses,
- ▶ new type annotations are inserted around case scrutinees,
- existing type annotations are normalized.

Algorithm Z, in one slide

This is an instance of the rule that deals with clauses:

...
$$a = a_1 \times a_2$$
, $(\Gamma; a : a_1; b : a_2) \vdash e \downarrow a_1 \times a_2 \rightsquigarrow e'$
true, $\Gamma \vdash (Pair \ a_1 \ a_2 \ a \ b : term \ a).e \downarrow a$
 $\rightsquigarrow (Pair \ a_1 \ a_2 \ a \ b).(e' : (a_1 \times a_2 \ b \ a))$

The clause is expected to return a value of type a. The equation $a=a_1\times a_2$ is available inside it. The body e is examined with the normalized expected shape $a_1\times a_2$. We insert an explicit coercion to let MLGX know about the equation that we are exploiting.

Programming in MLGX

This explains roughly how the surface language version of eval is transformed into:

```
\mu^*(eval: \forall a. \text{term } a \rightarrow a).\lambda t.
case\ (t: \text{term } a)\ of
|\ \text{Lit}\ i \rightarrow (i: (\text{int } \triangleright a))
|\ \text{Inc}\ t \rightarrow (eval\ t+1: (\text{int } \triangleright a))
|\ \text{Is}Z\ t \rightarrow (eval\ t=0: (bool\ \triangleright a))
|\ \text{If}\ b\ t\ e \rightarrow \text{if}\ eval\ b\ \text{then}\ eval\ t\ else}\ eval\ e
|\ \text{Pair}\ a_1\ a_2\ a\ b \rightarrow ((eval\ a, eval\ b): (a_1\times a_2\ \triangleright a))
|\ \text{Fst}\ a_2\ t \rightarrow \text{fst}\ (eval\ t)
|\ \text{Snd}\ a_1\ t \rightarrow \text{snd}\ (eval\ t)
```

Soundness

Theorem (Soundness for Algorithm Z)

Assume e has type σ in MLGI. If Z infers that e has shape s and rewrites e into e', then $s \leq \sigma$ holds and e' has type σ in MLGI.

The transformed program *can* be ill-typed in MLGX, but *never* because Z inserted incorrect annotations.

It's still unclear how relevant this theorem is in practice, but I like it.

Part III

Conclusion

Constraint-based type inference

Constraint-based type inference is a *versatile tool* that can deal with many language features while relying on a single constraint solver.

The solver's implementation can be complex, but its behavior remains predictable because it is correct and complete with respect to the logical interpretation of constraints.

Mandatory type annotations

Some constraint languages have intractable or undecidable satisfiability problems.

Instead of relying on an *incomplete* constraint solver, I suggest modifying the *constraint generation* process so as to take advantage of user-provided *hints*—typically, mandatory type annotations.

Stratified type inference

If the necessary hints are so numerous that they become a burden, a *local shape inference* algorithm can be used to automatically produce some of them.

Although its design is usually *ad hoc*, it should remain predictable if it is sufficiently *simple*.

Thank you.

Some questions

- ▶ is stratified type inference the way of the future, or a pis aller?
- ▶ is local shape inference really predictable?
- how do we explain type errors in a stratified system?
- ► can we allow some inferred type information to be fed back into shape inference, without losing predictability?

Selected References

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