

# 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  $\lambda$ -calculus

Hindley and Milner's type system

## Specification

The simply-typed  $\lambda$ -calculus is specified using a set of rules that allow deriving *judgements*:

$$\begin{array}{c}
 \text{Var} \\
 \Gamma \vdash x : \Gamma(x)
 \end{array}
 \qquad
 \begin{array}{c}
 \text{Abs} \\
 \frac{\Gamma; x : \tau_1 \vdash e : \tau_2}{\Gamma \vdash \lambda x. e : \tau_1 \rightarrow \tau_2}
 \end{array}
 \qquad
 \begin{array}{c}
 \text{App} \\
 \frac{\Gamma \vdash e_1 : \tau_1 \rightarrow \tau_2 \quad \Gamma \vdash e_2 : \tau_1}{\Gamma \vdash e_1 e_2 : \tau_2}
 \end{array}$$

The specification is *syntax-directed*.

## Substitutions versus constraints

Traditional presentations of type inference are based on substitutions, which means working with *most general unifiers*, *composition*, and *restriction*.

Reasoning in terms of constraints means working with *equations*, *conjunction*, and *existential quantification*.

Let's use the latter.

## 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  $\phi$  *satisfies* a constraint  $C$ .

Constraint solving is *first-order unification*.

## Constraint generation

Type inference is reduced to constraint solving by defining a mapping  $\llbracket \cdot \rrbracket$  of *pre-judgements* to constraints.

$$\begin{aligned} \llbracket \Gamma \vdash x : \tau \rrbracket &= \Gamma(x) = \tau \\ \llbracket \Gamma \vdash \lambda x. e : \tau \rrbracket &= \exists a_1 a_2. (\llbracket \Gamma; x : a_1 \vdash e : a_2 \rrbracket \wedge a_1 \rightarrow a_2 = \tau) \\ \llbracket \Gamma \vdash e_1 e_2 : \tau \rrbracket &= \exists a. (\llbracket \Gamma \vdash e_1 : a \rightarrow \tau \rrbracket \wedge \llbracket \Gamma \vdash e_2 : a \rrbracket) \end{aligned}$$

## Constraints, revisited

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

Let's enrich the syntax of constraints:

$$C ::= \dots \mid x = \tau \mid \text{def } x : \tau \text{ in } C$$

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

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

The *def* form is an *explicit substitution* form.

## Constraint generation, revisited

Constraint generation is now a mapping of an expression  $e$  and a type  $\tau$  to a constraint  $\llbracket e : \tau \rrbracket$ .

$$\begin{aligned} \llbracket x : \tau \rrbracket &= x = \tau \\ \llbracket \lambda x. e : \tau \rrbracket &= \exists a_1 a_2. \left( \begin{array}{l} \text{def } x : a_1 \text{ in } \llbracket e : a_2 \rrbracket \\ a_1 \rightarrow a_2 = \tau \end{array} \right) \\ \llbracket e_1 e_2 : \tau \rrbracket &= \exists a. (\llbracket e_1 : a \rightarrow \tau \rrbracket \wedge \llbracket e_2 : a \rrbracket) \end{aligned}$$

Look ma, *no environments!*

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

The simply-typed  $\lambda$ -calculus

Hindley and Milner's type system



## Specification

Three new typing rules are introduced *in addition* to those of the simply-typed  $\lambda$ -calculus:

$$\text{Gen} \quad \frac{\Gamma \vdash e : \tau \quad \bar{a} \# \text{ftv}(\Gamma)}{\Gamma \vdash e : \forall \bar{a}. \tau}$$

$$\text{Inst} \quad \frac{\Gamma \vdash e : \forall \bar{a}. \tau}{\Gamma \vdash e : [\vec{c}/\vec{a}]\tau}$$

$$\text{Let} \quad \frac{\Gamma \vdash e_1 : \sigma \quad \Gamma; x : \sigma \vdash e_2 : \tau}{\Gamma \vdash \text{let } x = e_1 \text{ in } e_2 : \tau}$$

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:

$$\begin{aligned}
 C &::= \tau = \tau \mid C \wedge C \mid \exists a.C \mid x \preceq \tau \mid \text{def } x : \zeta \text{ in } C \\
 \zeta &::= \forall \bar{a}[C].\tau
 \end{aligned}$$

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

$$\begin{aligned}
 \text{def } x : \zeta \text{ in } C &\equiv [\zeta/x]C \\
 (\forall \bar{a}[C].\tau) \preceq \tau' &\equiv \exists \bar{a}.(C \wedge \tau = \tau')
 \end{aligned}$$

## Constraint generation

Constraint generation is modified as follows:

$$\llbracket x : \tau \rrbracket = x \preceq \tau$$

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

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

## Statement

### Theorem (Soundness and completeness)

Let  $\Gamma$  be an environment whose domain is  $\text{fv}(e)$ . The expression  $e$  is well-typed relative to  $\Gamma$  iff

$$\text{def } \Gamma \text{ in } \exists a. \llbracket e : a \rrbracket$$

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.

$Lit :: \text{int} \rightarrow \text{term } \text{int}$

$Inc :: \text{term } \text{int} \rightarrow \text{term } \text{int}$

$IsZ :: \text{term } \text{int} \rightarrow \text{term } \text{bool}$

$If :: \forall a. \text{term } \text{bool} \rightarrow \text{term } a \rightarrow \text{term } a \rightarrow \text{term } a$

$Pair :: \forall a \beta. \text{term } a \rightarrow \text{term } \beta \rightarrow \text{term } (a \times \beta)$

$Fst :: \forall a \beta. \text{term } (a \times \beta) \rightarrow \text{term } a$

$Snd :: \forall a \beta. \text{term } (a \times \beta) \rightarrow \text{term } \beta$

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*:

$$\mu(\text{eval} : \forall a. \text{term } a \rightarrow a). \lambda t. \\ \text{case } t \text{ of}$$

- |  $\text{Lit } i \rightarrow (* a = \text{int } *) i$
- |  $\text{Inc } t \rightarrow (* a = \text{int } *) \text{eval } t + 1$
- |  $\text{IsZ } t \rightarrow (* a = \text{bool } *) \text{eval } t = 0$
- |  $\text{If } b \ t \ e \rightarrow \text{if } \text{eval } b \text{ then } \text{eval } t \text{ else } \text{eval } e$
- |  $\text{Pair } a \ b \rightarrow (* \exists a_1 a_2. a = a_1 \times a_2 *) (\text{eval } a, \text{eval } b)$
- |  $\text{Fst } t \rightarrow \text{fst } (\text{eval } t)$
- |  $\text{Snd } t \rightarrow \text{snd } (\text{eval } t)$

## 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:

$$\begin{aligned} \forall a\beta.\text{eq } a \beta \rightarrow a \rightarrow a \\ \forall a\beta.\text{eq } a \beta \rightarrow a \rightarrow \beta \\ \forall \gamma.\text{eq } \text{int } \text{bool} \rightarrow \text{int} \rightarrow \gamma \end{aligned}$$

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

$$\forall a\beta\gamma[a = \beta \Rightarrow a = \gamma].\text{eq } a \beta \rightarrow a \rightarrow \gamma$$

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 *g*eneralized algebraic data types in *i*mplicit 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*:

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

The symbol  $\Vdash$  stands for *constraint entailment*.

## Specification, continued

$$\frac{\text{Pair } a \ b : \text{term } a \vdash (a_1 a_2, a = a_1 \times a_2, a : a_1; b : a_2) \\ a_1 a_2 \# \text{ftv}(\Gamma, a) \quad a = a_1 \times a_2, (\Gamma; a : a_1; b : a_2) \vdash e : a}{\text{true}, \Gamma \vdash (\text{Pair } a \ b).e : \text{term } a \rightarrow a}$$

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...

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# 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

$$\frac{E, \Gamma \vdash (e : \text{term } a) : \text{term } a \quad \forall i \ E, \Gamma \vdash (p_i : \text{term } a). e_i : \text{term } a \rightarrow a}{E, \Gamma \vdash \text{case } (e : \text{term } a) \text{ of } p_1.e_1 \dots p_n.e_n : a}$$

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:

$$\frac{\text{Pair } a \ b : \text{term } a \vdash (a_1 a_2, a = a_1 \times a_2, a : a_1; b : a_2) \quad a_1 a_2 \# \text{ftv}(\Gamma, a) \quad a = a_1 \times a_2, (\Gamma; a : a_1; b : a_2) \vdash e : a}{\text{true}, \Gamma \vdash (\text{Pair } a \ b : \text{term } a).e : \text{term } a \rightarrow a}$$

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:

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

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 ( $\text{int} \triangleright a$ ) is just a constant of type  $\text{int} \rightarrow a$ .

*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:

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

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

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# Shapes

*Shapes* are defined by

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

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

That is, the shape  $\gamma.\gamma \rightarrow \gamma$  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,

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

The *uninformative* shape  $\gamma.\gamma$ , written  $\perp$ , 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,

$$(\gamma.\gamma \rightarrow \gamma) \sqcup (\gamma.\text{int} \rightarrow \gamma) = \text{int} \rightarrow \text{int}$$

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

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

where  $\Gamma$  (which maps variables to shapes) and  $s$  are *provided*, while  $s'$  is *inferred* and at least as informative, that is,  $s \preceq s'$  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:

$$\frac{\dots \quad a = a_1 \times a_2, (\Gamma; a : a_1; b : a_2) \vdash e \downarrow a_1 \times a_2 \rightsquigarrow e'}{\text{true}, \Gamma \vdash (\text{Pair } a_1 \ a_2 \ a \ b : \text{term } a). e \downarrow a \rightsquigarrow (\text{Pair } a_1 \ a_2 \ a \ b).(e' : (a_1 \times a_2 \triangleright a))}$$

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:

$$\mu^*(eval : \forall a. term\ a \rightarrow a). \lambda t.$$

case (t : term a) of

- | Lit i → (i : (int ▷ a))
- | Inc t → (eval t + 1 : (int ▷ a))
- | IsZ t → (eval t = 0 : (bool ▷ a))
- | If b t e → if eval b then eval t else eval e
- | Pair a<sub>1</sub> a<sub>2</sub> a b → ((eval a, eval b) : (a<sub>1</sub> × a<sub>2</sub> ▷ a))
- | Fst a<sub>2</sub> t → fst (eval t)
- | Snd a<sub>1</sub> t → snd (eval t)

# Soundness

## Theorem (Soundness for Algorithm Z)

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

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