MPRI, Typage

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Introduction

Simply-typed λ -calculus

Polymorphism and System F

Type reconstruction

Type reconstruction

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Introduction Simple types Core ML Type annotations Recursive Types System F Logical versus algorithmic properties

We have viewed a type system as a 3-place predicate over a type environment, a term, and a type.

So far, we have been concerned with *logical* properties of the type system, namely subject reduction and progress.

However, one should also study its *algorithmic* properties: is it decidable whether a term is well-typed?

Introduction Simple types Core ML Type annotations Recursive Types System F Logical versus algorithmic properties

We have seen three different type systems, simply-typed λ -calculus, ML, and System F, of increasing expressiveness.

In each case, we have presented an explicitly-typed and an implicitly-typed version of the language and show a close correspondence between the two views, thanks to a type-passing semantics.

We argued that the explicitly-typed version is often more convenient for studying the meta-theoretical properties of the language.

Which one should we used for checking well-typedness? That is, in which language should we write programs?

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Checking	type deri	vations			

The typing judgment is *inductively defined*, so that, in order to prove that a particular instance holds, one exhibits a type derivation.

A type derivation is essentially a version of the program where every node is annotated with a type.

Checking that a type derivation is correct is usually easy: it basically amounts to checking equalities between types.

However, type derivations are so verbose as to be intractable by humans! Requiring every node to be type-annotated is not practical.

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Bottom-u	p type-che	cking			

A more practical, and quite common, approach consists in requesting just enough annotations to allow types to be reconstructed in a *bottom-up* manner.

In other words, one seeks an algorithmic reading of the typing rules, where, in a judgment $\Gamma \vdash t:T$, the parameters Γ and t are inputs, while the parameter T is an output.

Moreover, typing rules should be such that a type appearing as output in a conclusion should also appear as output in a premise or as input in the conclusion and input in the premises should be input of the conclusion or output of other premises.

This way, types need never be guessed, just looked up into the typing context, instantiated, or checked for equality.

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Bottom	-up type-cl	hecking			

This is exactly the situation with explicitly-typed presentations of the typing rules.

This is also the traditional approach of Pascal, C, C++, Java, \dots : formal procedure parameters, as well as local variables, are assigned explicit types. The types of expressions are synthesized bottom-up.

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Bottom-	up type-c	hecking			

However, this implies a lot of redundancies:

- Parameters of *all* functions need to be annotated, even when their types are obvious from context.
- Let-expressions (when not primitive), recursive definitions, Injection into sum types need to be annotated.
- As the language grows, more and more constructs require type annotations, *e.g.* type applications and type abstractions.

Type annotations may quickly obfuscate the code and large explicitly-typed terms are so verbose that they become intractable by humans!

Hence, programming in the implicitly-typed version is more appealing.

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Type inf	erence				

For simply-typed λ -calculus and ML, it turns out that this is possible: whether a term is well-typed is decidable, even when no type annotations are provided!

For System F, this is however undecidable. Since programming in explicitly-typed System F is not practically feasible, some amount of type reconstruction must still be done. Typically, the algorithm is incomplete, *i.e.* it rejects terms that are perhaps well-typed, but the user may always provide more annotations and at worse, the explicitly-typed version is never rejected if well-typed.

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Type inf	erence				

The type inference algorithm for simply-typed λ -calculus, is due to Hindley [1969]. The idea behind the algorithm is simple.

Because simply-typed λ -calculus is a syntax-directed type system, an unannotated term determines an isomorphic candidate type derivation, where all types are unknown: they are distinct type variables.

For a candidate type derivation to become an actual, valid type derivation, every type variable must be instantiated with a type, subject to certain *equality constraints* on types.

For instance, at an application node, the type of the operator must match the domain type of the operator.

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Type inf	erence				

Thus, type inference for the simply-typed λ -calculus decomposes into constraint generation followed by constraint solving.

Simple types are first-order terms. Thus, solving a collection of equations between simple types is first-order unification.

First-order unification can be performed incrementally in quasi-linear time, and admits particularly simple *solved forms*.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Constra	ints				

At the interface between the constraint generation and constraint solving phases is the *constraint language*.

It is a logic: a syntax, equipped with an interpretation in a model.

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Constra	ints				

There are two syntactic categories: types and constraints.

A type is either a type variable X or an arity-consistent application of a type constructor F.

(The type constructors are unit, \times , +, \rightarrow , etc.)

An atomic constraint is truth, falsity, or an equation between types.

Compound constraints are built on top of atomic constraints via *conjunction* and *existential quantification* over type variables.

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Constra	ints				

Constraints are interpreted in the Herbrand universe, that is, in the set of ground types:

t ::= F t

Ground types contain no variables. The base case in this definition is when F has arity zero.

A ground assignment $\boldsymbol{\phi}$ is a total mapping of type variables to ground types.

A ground assignment determines a total mapping of types to ground types.

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Constrai	nts				

The interpretation of constraints takes the form of a judgment, $\phi \vdash C$, pronounced: ϕ satisfies C, or ϕ is a solution of C.

This judgment is inductively defined:

$$\phi \vdash \text{true} \qquad \frac{\phi T_1 = \phi T_2}{\phi \vdash T_1 = T_2} \qquad \frac{\phi \vdash C_1 \quad \phi \vdash C_2}{\phi \vdash C_1 \land C_2} \qquad \frac{\phi [X \mapsto \mathbf{t}] \vdash C}{\phi \vdash \exists X.C}$$

A constraint C is satisfiable if and only if there exists a ground assignment ϕ that satisfies C.

We write $C_1 \equiv C_2$ when C_1 and C_2 have the same solutions.

The problem: "given a constraint C, is C satisfiable?" is first-order unification.

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Constrai	nt genera	tion			

Type inference is reduced to constraint solving by defining a mapping of *candidate judgments* to constraints.

$$\begin{bmatrix} \Gamma \vdash x : T \end{bmatrix} = \Gamma(x) = T$$
$$\begin{bmatrix} \Gamma \vdash \lambda x.a : T \end{bmatrix} = \exists X_1 X_2.(\llbracket \Gamma, x : X_1 \vdash a : X_2 \rrbracket \land X_1 \to X_2 = T)$$
$$if X_1, X_2 \# \Gamma, a, T$$
$$\begin{bmatrix} \Gamma \vdash a_1 a_2 : T \end{bmatrix} = \exists X.(\llbracket \Gamma \vdash a_1 : X \to T \rrbracket \land \llbracket \Gamma \vdash a_2 : X \rrbracket)$$
$$if X \# \Gamma, a_1, a_2, T$$

Thanks to the use of existential quantification, the names that occur free in $[\Gamma \vdash a:T]$ are a subset of those that occur free in Γ or T.

This allows the freshness side-conditions to remain *local* — there is no need to informally require "globally fresh" type variables.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
An exan	nple				

Let us perform type inference for the closed term

 $\lambda f x y. (f x, f y)$

The problem is to construct and solve the constraint

 $\llbracket \emptyset \vdash \lambda f x y. (f x, f y) : X_0 \rrbracket$

It is possible (and, for a human, easier) to mix these tasks. A machine, however, could generate and solve in two successive phases. Solving the constraint means to find all possible ground assignments for X_0 that satisfy the constraint.

Typically, this is done by transforming the constraint into successive equivalent constraints until some constraint that is obviously satisfiable and from which solutions may be directly read.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
An exar	nple				

$$\begin{split} & \left[\varnothing \vdash \lambda f_{Xy}. \left(f_{X}, f_{y} \right) : X_{0} \right] \\ &= \exists X_{1} X_{2}. \left(\begin{array}{c} \left[f : X_{1} \vdash \lambda xy. \ldots : X_{2} \right] \\ X_{1} \rightarrow X_{2} = X_{0} \end{array} \right) \\ &= \exists X_{1} X_{2}. \left(\begin{array}{c} \exists X_{3} X_{4}. \left(\begin{array}{c} \left[f : X_{1}; x : X_{3} \vdash \lambda y. \ldots : X_{4} \right] \\ X_{3} \rightarrow X_{4} = X_{2} \end{array} \right) \\ X_{1} \rightarrow X_{2} = X_{0} \end{array} \right) \\ &= \exists X_{1} X_{2}. \left(\begin{array}{c} \exists X_{3} X_{4}. \left(\begin{array}{c} \left[f : X_{1}; x : X_{3} \vdash \lambda y. \ldots : X_{4} \right] \\ X_{3} \rightarrow X_{4} = X_{2} \end{array} \right) \\ X_{1} \rightarrow X_{2} = X_{0} \end{array} \right) \\ &= \exists X_{1} X_{2}. \left(\begin{array}{c} \exists X_{3} X_{4}. \left(\begin{array}{c} \exists X_{5} X_{6}. \left(\left[f : X_{1}; x : X_{3}; y : X_{5} \vdash (f \times, f \cdot y) : X_{6} \right] \\ X_{3} \rightarrow X_{4} = X_{2} \end{array} \right) \\ X_{1} \rightarrow X_{2} = X_{0} \end{array} \right) \end{split}$$

We have performed constraint generation for the 3 $\lambda\text{-abstractions.}$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
An exan	ıple				

$$\exists X_1 X_2. \left(\begin{array}{c} \exists X_3 X_4. \begin{pmatrix} \exists X_5 X_6. \begin{pmatrix} \llbracket f: X_1; x: X_3; y: X_5 \vdash (f x, f y): X_6 \rrbracket \\ X_5 \to X_6 = X_4 \end{pmatrix} \\ X_1 \to X_2 = X_0 \end{array} \right) \right)$$

$$\equiv \exists X_{1}X_{2}X_{3}X_{4}X_{5}X_{6}. \begin{pmatrix} [[f:X_{1};x:X_{3};y:X_{5} \vdash (fx,fy):X_{6}]] \\ X_{5} \to X_{6} = X_{4} \\ X_{3} \to X_{4} = X_{2} \\ X_{1} \to X_{2} = X_{0} \end{pmatrix}$$

We have hoisted up several existential quantifiers:

 $(\exists X.C_1) \land C_2 \equiv \exists X.(C_1 \land C_2) \quad \text{if } X \# C_2$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
An exan	nple				

$$\exists X_{1}X_{2}X_{3}X_{4}X_{5}X_{6}. \begin{pmatrix} [[f:X_{1};x:X_{3};y:X_{5} \vdash (fx,fy):X_{6}]] \\ X_{5} \rightarrow X_{6} = X_{4} \\ X_{3} \rightarrow X_{4} = X_{2} \\ X_{1} \rightarrow X_{2} = X_{0} \end{pmatrix}$$

$$= \exists X_1 X_2 X_3 X_5 X_6. \left(\begin{array}{c} [[f:X_1; x:X_3; y:X_5 \vdash (f x, f y):X_6]] \\ X_3 \to X_5 \to X_6 = X_2 \\ X_1 \to X_2 = X_0 \end{array} \right)$$

We have eliminated a type variable (X_4) with a defining equation: $\exists X. (C \land X = T) \equiv [X \mapsto T]C \quad \text{if } X \# T$

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An exan	ıple				

$$\exists X_1 X_2 X_3 X_5 X_6. \left(\begin{array}{c} \left[f: X_1; x: X_3; y: X_5 \vdash (f \, x, f \, y) : X_6 \right] \\ X_3 \to X_5 \to X_6 = X_2 \\ X_1 \to X_2 = X_0 \end{array} \right)$$

$$\equiv \exists X_1 X_3 X_5 X_6. \left(\begin{array}{c} \llbracket f : X_1; x : X_3; y : X_5 \vdash (f x, f y) : X_6 \rrbracket \\ X_1 \to X_3 \to X_5 \to X_6 = X_0 \end{array} \right)$$

We have again eliminated a type variable (X_2) with a defining equation. In the following, let Γ stand for $(f: X_1; x: X_3; y: X_5)$.

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An exar	nple				

$$\exists X_{1}X_{3}X_{5}X_{6}.\left(\begin{array}{c} \left[\left[\Gamma \vdash (f \times, f y) : X_{6}\right]\right] \\ X_{1} \rightarrow X_{3} \rightarrow X_{5} \rightarrow X_{6} = X_{0} \end{array}\right)$$
$$\equiv \exists X_{1}X_{3}X_{5}X_{6}X_{7}X_{8}.\left(\begin{array}{c} \left[\left[\Gamma \vdash f \times : X_{7}\right]\right] \\ \left[\left[\Gamma \vdash f y : X_{8}\right]\right] \\ X_{7} \times X_{8} = X_{6} \\ X_{1} \rightarrow X_{3} \rightarrow X_{5} \rightarrow X_{6} = X_{0} \end{array}\right)$$

$$\equiv \exists X_1 X_3 X_5 X_7 X_{\mathcal{B}}. \left(\begin{array}{c} \left[\Gamma \vdash f \times : X_7 \right] \\ \left[\Gamma \vdash f y : X_{\mathcal{B}} \right] \\ X_1 \to X_3 \to X_5 \to X_7 \times X_{\mathcal{B}} = X_0 \end{array} \right)$$

We have performed constraint generation for the pair, hoisted the resulting existential quantifiers, and eliminated a type variable (X_6) . Let us now focus on the first application...

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An exam	ıple				

$$\begin{bmatrix} \Gamma \vdash f \times : X_7 \end{bmatrix}$$

= $\exists X_9. \left(\begin{bmatrix} \Gamma \vdash f : X_9 \rightarrow X_7 \end{bmatrix} \right)$
= $\exists X_9. \left(X_1 = X_9 \rightarrow X_7 \end{bmatrix}$
 $X_3 = X_9$

$$\equiv X_1 = X_3 \rightarrow X_7$$

We have performed constraint generation for the variables f and x, and eliminated a type variable (X_9).

Recall that Γ stands for $(f:X_1; x:X_3; y:X_5)$.

Now, back to the big picture ...

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An exan	nple				

$$\exists X_1 X_3 X_5 X_7 X_{\mathcal{B}}. \left(\begin{array}{c} \left[\left[\Gamma \vdash f \times : X_7 \right] \right] \\ \left[\left[\Gamma \vdash f y : X_{\mathcal{B}} \right] \right] \\ X_1 \to X_3 \to X_5 \to X_7 \times X_{\mathcal{B}} = X_0 \end{array} \right)$$

$$\equiv \exists X_1 X_3 X_5 X_7 X_{\mathcal{B}}. \left(\begin{array}{c} X_1 = X_3 \to X_7 \\ \llbracket \Gamma \vdash f \ y : X_{\mathcal{B}} \rrbracket \\ X_1 \to X_3 \to X_5 \to X_7 \times X_{\mathcal{B}} = X_0 \end{array} \right)$$

$$= \exists X_1 X_3 X_5 X_7 X_8. \begin{pmatrix} X_1 = X_3 \rightarrow X_7 \\ X_1 = X_5 \rightarrow X_8 \\ X_1 \rightarrow X_3 \rightarrow X_5 \rightarrow X_7 \times X_8 = X_0 \end{pmatrix}$$

We have applied the previous simplification under a context:

$$C_1 \equiv C_2 \Rightarrow C[C_1] \equiv C[C_2]$$

We have simplified the right-hand application analogously.

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An exan	nple				

$$\exists X_1 X_3 X_5 X_7 X_8. \left(\begin{array}{c} X_1 = X_3 \to X_7 \\ X_1 = X_5 \to X_8 \\ X_1 \to X_3 \to X_5 \to X_7 \times X_8 = X_0 \end{array}\right)$$

$$\equiv \exists X_1 X_3 X_5 X_7 X_{\mathcal{B}}. \begin{pmatrix} X_1 = X_3 \to X_7 \\ X_3 = X_5 \\ X_7 = X_8 \\ X_1 \to X_3 \to X_5 \to X_7 \times X_8 = X_0 \end{pmatrix}$$

$$\equiv \exists X_3 X_7. \left((X_3 \to X_7) \to X_3 \to X_3 \to X_7 \times X_7 = X_0 \right)$$

We have applied transitivity at X_1 , structural decomposition, and eliminated three type variables (X_1 , X_5 , X_8).

We have now reached a solved form.

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An exan	nple				

We have checked the following equivalence:

$$\begin{bmatrix} \emptyset \vdash \lambda f \times y. (f \times, f y) : X_0 \end{bmatrix}$$

= $\exists X_3 X_7. ((X_3 \rightarrow X_7) \rightarrow X_3 \rightarrow X_3 \rightarrow X_7 \times X_7 = X_0)$

The ground types of λf_{xy} . (f x, f y) are all ground types of the form $(\mathbf{t}_3 \rightarrow \mathbf{t}_7) \rightarrow \mathbf{t}_3 \rightarrow \mathbf{t}_3 \rightarrow \mathbf{t}_7 \times \mathbf{t}_7$.

 $(X_3 \rightarrow X_7) \rightarrow X_3 \rightarrow X_3 \rightarrow X_7 \times X_7$ is a principal type for $\lambda fxy.(fx, fy)$.

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An exan	nple				

Objective Caml implements a form of this type inference algorithm:

This technique is used also by Standard ML and Haskell.

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An exan	ıple				

In the simply-typed λ -calculus, type inference works just as well for *open* terms. Consider, for instance:

 $\lambda x y. (f x, f y)$

This term has a free variable, namely f.

The type inference problem is to construct and solve the constraint

 $\llbracket f: X_1 \vdash \lambda xy. (f x, f y) : X_2 \rrbracket$

We have already done so... with only a slight difference: X_1 and X_2 are now free, so they cannot be eliminated.

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An exan	nple				

One can check the following equivalence:

$$\begin{bmatrix} f: X_1 \vdash \lambda x y. (f x, f y) : X_2 \end{bmatrix}$$

= $\exists X_3 X_7. \begin{pmatrix} X_3 \rightarrow X_7 = X_1 \\ X_3 \rightarrow X_3 \rightarrow X_7 \times X_7 = X_2 \end{pmatrix}$

In other words, the ground typings of $\lambda xy.(f x, f y)$ are all ground typings of the form:

$$((f: \mathbf{t}_3 \rightarrow \mathbf{t}_7), \mathbf{t}_3 \rightarrow \mathbf{t}_3 \rightarrow \mathbf{t}_7 \times \mathbf{t}_7)$$

Remember that a typing is a pair of an environment and a type.

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Typings					

Definition

 (Γ, T) is a typing of a if and only if dom $(\Gamma) = fv(a)$ and the judgment $\Gamma \vdash a:T$ is valid.

The type inference problem is to determine whether a term a admits a typing, and, if possible, to exhibit a description of the set of all of its typings.

Up to a change of universes, the problem reduces to finding the *ground typings* of a term. (For every type variable, introduce a nullary type constructor. Then, ground typings in the extended universe are in one-to-one correspondence with typings in the original universe.)

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Constrai	int genera	tion			

Theorem (Soundness and completeness)

 $\phi \vdash \llbracket \Gamma \vdash a : T \rrbracket$ if and only if $\phi \Gamma \vdash a : \phi T$.

Proof.

By structural induction over a. (Recommended exercise.)

In other words, assuming $dom(\Gamma) = fv(a)$, ϕ satisfies the constraint $[\Gamma \vdash a:T]$ if and only if $(\phi\Gamma, \phiT)$ is a (ground) typing of a.

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Constrai	int genera	tion			

Corollary

Let $fv(a) = \{x_1, \dots, x_n\}$, where $n \ge 0$. Let X_0, \dots, X_n be pairwise distinct type variables. Then, the ground typings of a are described by $((x_i : \phi X_i)_{i \le 1..n}, \phi X_0)$

where ϕ ranges over all solutions of $[(x_i : X_i)_{i \in 1..n} \vdash a : X_0]]$.

Corollary

Let $fv(a) = \emptyset$. Then, a is well-typed if and only if $\exists X. [\![\emptyset \vdash a : X]\!] \equiv true$.

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Constra	int solving				

A constraint solving algorithm is typically presented as a (non-deterministic) system of *constraint rewriting rules*.

The system must enjoy the following properties:

- reduction is meaning-preserving: $C_1 \longrightarrow C_2$ implies $C_1 \equiv C_2$;
- reduction is terminating;
- every normal form is either "false" (literally) or satisfiable.

The normal forms are called solved forms.

Following Pottier and Rémy [2005, §10.6], we extend the syntax of constraints and replace ordinary binary equations with *multi-equations*:

 $U ::= true | false | \epsilon | U \land U | \exists \overline{X}.U$

A multi-equation ϵ is a multi-set of types. Its interpretation is:

$$\frac{\forall T \in e, \quad \phi T = \mathbf{t}}{\phi \vdash e}$$

That is, ϕ satisfies ϵ if and only if ϕ maps all members of ϵ to a single ground type.

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First-ord	der unificat	tion as (constraint	solving	

-

See [Pottier and Rémy, 2005, $\S10.6]$ for additional administrative rules.

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The occ	urs check				

X dominates Y (with respect to U) iff U contains a multi-equation of the form $FT_1 \dots Y \dots T_n = X = \dots$

U is cyclic iff its domination relation is cyclic.

A cyclic constraint is unsatisfiable: indeed, if ϕ satisfies U and if X is a member of a cycle, then the ground type ϕX must be a strict subterm of itself, a contradiction.

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Solved	forms				

A solved form is either false or $\exists \bar{X}.U$, where U is a conjunction of multi-equations, every multi-equation contains at most one non-variable term, no two multi-equations share a variable, and the domination relation is acyclic.

Every solved form of that is not false is satisfiable — indeed, a solution is easily constructed by well-founded recursion over the domination relation.

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Impleme	ntation				

Viewing a unification algorithm as a system of rewriting rules makes it easy to explain and reason about.

In practice, following Huet [1976], first-order unification is implemented on top of an efficient *union-find* data structure [Tarjan, 1975]. Its time complexity is quasi-linear.

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Closing	remarks				

Thanks to type inference, *conciseness* and *static safety* are not incompatible.

Furthermore, an inferred type is sometimes *more general* than a programmer-intended type. Type inference helps reveal unexpected generality.

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Two pre	sentations				

Two presentations of type inference for Damas and Milner's type system are possible:

- one of Milner's classic algorithms [1978], \mathcal{W} or \mathcal{J} ; see Pottier's old course notes for details [Pottier, 2002, §3.3];
- a constraint-based presentation [Pottier and Rémy, 2005];

We favor the latter, but quickly review the former first.

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Prelimina	aries				

This algorithm expects a pair $\Gamma \vdash a$, produces a type T, and uses two global variables, V and φ .

V is an infinite fresh supply of type variables:

$$fresh = do X \in V$$
$$do V \leftarrow V \setminus \{X\}$$
$$return X$$

 φ is an idempotent substitution (of types for type variables), initially the identity.

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The algo	orithm				

Here is the algorithm in monadic style:

$$\mathcal{J}(\Gamma \vdash x) = \det \forall X_1 \dots X_n. T = \Gamma(x) do X'_1, \dots, X'_n = \text{fresh}, \dots, \text{fresh} return [X_i \mapsto X'_i]_{i=1}^n (T) - take a fresh instance $\mathcal{J}(\Gamma \vdash \lambda x. a_1) = do X = \text{fresh} do T_1 = \mathcal{J}(\Gamma; x : X \vdash a_1) return X \to T_1 - form an arrow type \dots$$$

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The algo	prithm				

$$\mathcal{J}(\Gamma \vdash a_1 a_2) = do T_1 = \mathcal{J}(\Gamma \vdash a_1) do T_2 = \mathcal{J}(\Gamma \vdash a_2) do X = fresh do \varphi \leftarrow mgu(\varphi(T_1) = \varphi(T_2 \rightarrow X)) \circ \varphi return X - solve T_1 = T_2 \rightarrow X \mathcal{J}(\Gamma \vdash let x = a_1 in a_2) = do T_1 = \mathcal{J}(\Gamma \vdash a_1) let \sigma = \forall \setminus ftv(\varphi(\Gamma)). \varphi(T_1) - generalize return \mathcal{J}(\Gamma; x : \sigma \vdash a_2)$$

 $(\forall \setminus \bar{X}.T \text{ quantifies over all type variables other than } \bar{X}.)$

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Some wa	eaknesses				

Algorithm \mathcal{J} mixes generation and solving of equations. This lack of modularity leads to several weaknesses:

- proofs are more difficult;
- correctness and efficiency concerns are not clearly separated (if implemented literally, the algorithm is exponential in practice);
- adding new language constructs duplicates solving of equations;
- generalizations, such as the introduction of subtyping, are not easy.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Some v	veaknesses				

Algorithm ${\mathcal J}$ works with substitutions, instead of constraints.

Substitutions are an approximation to solved forms for unification constraints.

Working with substitutions means using most general unifiers, composition, and restriction.

Working with constraints means using equations, conjunction, and existential quantification.

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- \bullet Introduction
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Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Road m	ар				

Type inference for Damas and Milner's type system involves slightly more than first-order unification: there is also *generalization* and *instantiation* of type schemes.

So, the constraint language must be enriched.

We proceed in two steps:

- still within simply-typed λ -calculus, we present a variation of the constraint language;
- building on this variation, we introduce polymorphism.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
A variat	cion on cor	istraints			

How about letting the constraint solver, instead of the constraint generator, deal with *environment access* and *construction?*

Let's enrich the syntax of constraints:

$$C ::= \dots | x = T | def x : T in C$$

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

$$def x : T in C \equiv [x \mapsto T]C$$

The def form is an *explicit substitution* form.

More precisely, here is the new interpretation of constraints. As before, a valuation ϕ maps type variables X to ground types. In addition, a valuation ψ maps term variables x to ground types. The satisfaction judgment now takes the form $\phi, \psi \vdash C$. The new rules of interest are:

$$\frac{\psi x = \phi T}{\phi, \psi \vdash x = T} \qquad \qquad \frac{\phi, \psi [x \mapsto \phi T] \vdash C}{\phi, \psi \vdash def \ x : T \ in \ C}$$

(All other rules are modified to just transport ψ .)

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
A variat	ion on col	nstraints			

Constraint generation is now a mapping of an expression a and a type T to a constraint [a:T]. There is no longer a need for the parameter Γ .

$$\begin{bmatrix} x : T \end{bmatrix} = x = T$$

$$\begin{bmatrix} \lambda x. a : T \end{bmatrix} = \exists X_1 X_2. (def x : X_1 in [[a : X_2]] \land X_1 \to X_2 = T)$$

if $X_1, X_2 # a, T$

$$\begin{bmatrix} a_1 a_2 : T \end{bmatrix} = \exists X. ([[a_1 : X \to T]] \land [[a_2 : X]])$$

if $X # a_1, a_2, T$

Look ma, no environments!

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
A variat	ion on coi	nstraints			

Theorem (Soundness and completeness)

Assume $fv(a) = dom(\Gamma)$. Then, $\phi, \phi\Gamma \vdash [a:T]]$ if and only if $\phi\Gamma \vdash a:\phiT$.

Corollary

Assume $fv(a) = \emptyset$. Then, a is well-typed if and only if $\exists X.[[a:X]] \equiv true$.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Summary	/				

This variation shows that there is *freedom* in the design of the constraint language, and that altering this design can *shift work* from the constraint generator to the constraint solver, or vice-versa.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Enriching	constraint	55			

To permit polymorphism, we must extend the syntax of constraints so that a variable x denotes not just a ground type, but a set of ground types.

However, these sets cannot be represented as type schemes $\forall \bar{X}.T$, because constructing these simplified forms requires constraint solving.

To avoid mingling constraint generation and constraint solving, we use type schemes that incorporate constraints: *constrained type schemes*.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Enriching	constrair	nts			

The syntax of constraints and of constrained type schemes is:

$$C ::= T = T | C \land C | \exists X.C$$
$$| x \leq T$$
$$| \sigma \leq T$$
$$| def x : \sigma in C$$
$$\sigma ::= \forall \overline{X}[C].T$$

 $x \leq T$ and $\sigma \leq T$ are *instantiation constraints*. The latter form is introduced so as to make the syntax stable under substitutions of constrained type schemes for variables.

As before, def $x : \sigma$ in C is an explicit substitution form.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Enriching	constraint	ts			

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

$$def x : \sigma \text{ in } C \equiv [x \mapsto \sigma]C$$
$$(\forall \bar{X}[C].T) \leq T' \equiv \exists \bar{X}.(C \land T = T') \quad \text{if } \bar{X} \# T'$$

Using these laws, a closed constraint can be rewritten to a unification constraint (with a possibly exponential increase in size).

The new constructs do not add much expressive power. They add just enough to allow a stand-alone formulation of constraint generation.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Interpretir	ng consti	raints			

A type variable X still denotes a ground type.

A variable x now denotes a set of ground types.

Instantiation constraints are interpreted as set membership.

$$\frac{\phi T \in \psi x}{\phi, \psi \vdash x \leq T} \qquad \frac{\phi T \in \binom{\phi}{\psi} \sigma}{\phi, \psi \vdash \sigma \leq T} \qquad \frac{\phi, \psi [x \mapsto \binom{\phi}{\psi} \sigma] \vdash C}{\phi, \psi \vdash def \ x : \sigma \text{ in } C}$$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Interpret	ting const	rained tv	ype schemes	5	

The interpretation of $\forall \bar{X}[C]$. T under ϕ and ψ is the set of all $\phi'T$, where ϕ and ϕ' coincide outside \bar{X} and where ϕ' and ψ satisfy C.

$$\begin{pmatrix} \phi \\ \psi \end{pmatrix} (\forall \bar{X}[C].\mathcal{T}) = \{ \phi' \mathcal{T} \mid (\phi' \smallsetminus \bar{X} = \phi \smallsetminus \bar{X}) \land (\phi', \psi \vdash C) \}$$

For instance, the interpretation of $\forall X [\exists Y.X = Y \rightarrow Z]. X \rightarrow X$ under ϕ and ψ is the set of all ground types of the form $(t \rightarrow \phi Z) \rightarrow (t \rightarrow \phi Z)$, where t ranges over ground types. This is also the interpretation of $\forall Y. (Y \rightarrow Z) \rightarrow (Y \rightarrow Z)$. In fact, every constrained type scheme is equivalent to a standard type scheme.

If \bar{X} and C are empty, then $\begin{pmatrix} \phi \\ \psi \end{pmatrix} T$ is ϕT .

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
A derived	d form				

Notice that if x does not appear free in C, def $x:\sigma$ in C is equivalent to C—whether or not of the constraints appearing in σ are solvable. To enforce the constraints in σ to be solvable, we use a variant of the *def* construct:

let $x : \sigma$ in $C \equiv def x : \sigma$ in $((\exists X.x \leq X) \land C)$

Expanding $\sigma \stackrel{\Delta}{=} \forall \bar{X}[C]$. \mathcal{T} and simplifying, an equivalent definition is: let $x : \forall \bar{X}[C]$. \mathcal{T} in $C' \equiv \exists \bar{X}. C \land def x : \forall \bar{X}[C]$. \mathcal{T} in C'

It would also be equivalent to provide a direct interpretation of it:

$$\frac{\begin{pmatrix} \phi \\ \psi \end{pmatrix} \sigma \neq \emptyset \qquad \phi, \psi[x \mapsto \begin{pmatrix} \phi \\ \psi \end{pmatrix} \sigma] \vdash C}{\phi, \psi \vdash \text{let } x : \sigma \text{ in } C}$$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Constrai	nt genera [.]	tion			

Constraint generation is now as follows:

let

$$\begin{bmatrix} x:T \end{bmatrix} = x \leq T$$

$$\begin{bmatrix} \lambda x.a:T \end{bmatrix} = \exists X_1 X_2. (\det x:X_1 \text{ in } \begin{bmatrix} a:X_2 \end{bmatrix} \land X_1 \rightarrow X_2 = T)$$

$$\text{if } X_1, X_2 \neq a, T$$

$$\begin{bmatrix} a_1 \ a_2:T \end{bmatrix} = \exists X. (\begin{bmatrix} a_1:X \rightarrow T \end{bmatrix} \land \begin{bmatrix} a_2:X \end{bmatrix})$$

$$\text{if } X \neq a_1, a_2, T$$

$$x = a_1 \text{ in } a_2:T \end{bmatrix} = \det x: (a_1) \text{ in } \begin{bmatrix} a_2:T \end{bmatrix}$$

$$(a) = \forall X \begin{bmatrix} \begin{bmatrix} a:X \end{bmatrix} \end{bmatrix}. X$$

(a) is a principal constrained type scheme for a: its intended interpretation is the set of all ground types that a admits.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Properti	ies of cons [.]	traint g	eneration		
Lemma ∃X.([[a : X	X] $\land X = T$) =	[[a:T]] in	f X # T.		
Lemma ((a)) ≤ T	≡ [[a:T]].				
Lemma [x ↦ (a ₁))][[a ₂ : <i>T</i>]] ≡	$\llbracket [x \mapsto a_1],$	a ₂ : T]].		
Lemma [[let x = a	a₁ in a₂ : T]] =	[[a₁; [x ⊢	≻ a ₁]a ₂ : T]].		

The constraint associated with a let construct is *equivalent* to the constraint associated with its let-normal form.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Complexi	ty				

Lemma

The size of [a:T] is linear in the sum of the sizes of a and T.

Constraint generation can be implemented in linear time and space.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Soundne	ess and co	ompletene	866		

The statement keeps its previous form, but Γ now contains Damas-Milner type schemes. Since Γ binds variables to type schemes, we define $\phi(\Gamma)$ as the point-wise mapping of $\binom{\phi}{\emptyset}$ to Γ .

Theorem (Soundness and completeness)

Let $fv(a) = dom(\Gamma)$. Then, $\phi, \phi \Gamma \vdash [a:T]$ if and only if $\phi \Gamma \vdash a:\phi T$.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
A word	about HM	(X)			

Soundness/completeness of type inference are in fact easier to prove if one adopts a constraint-based specification of the type system.

In HM(X), typing judgments take the form $C, \Gamma \vdash a: T$. The system includes a subtyping rule:

$$\frac{C, \Gamma \vdash a: T_1 \qquad C \Vdash T_1 \leq T_2}{C, \Gamma \vdash a: T_2}$$

This generalizes Damas and Milner's type system.

See Odersky et al. [1999], Pottier and Rémy [2005], Skalka and Pottier [2002].

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Summary					

Note that

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

This makes constraints suitable for use in an efficient and modular implementation.

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Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
An initial	environn	nent			

Let Γ_0 stand for assoc: $\forall XY. X \rightarrow \text{list}(X \times Y) \rightarrow Y$.

We take Γ_0 to be the initial environment, so that the constraints considered next are implicitly wrapped within the context def Γ_0 in [].

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
A code	fragment				

Let a stand for the term

```
\lambda x.\lambda l_1.\lambda l_2.
let assocx = assoc x in
(assocx l_1, assocx l_2)
```

One anticipates that assocx receives a polymorphic type scheme, which is instantiated twice at different types...

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
The gene	erated con	nstraint			

Let Γ stand for $x: X_0; l_1: X_1; l_2: X_2$. Then, the constraint [a:X] is (with a few minor simplifications):

$$\exists X_0 X_1 X_2 Y. \begin{pmatrix} X = X_0 \rightarrow X_1 \rightarrow X_2 \rightarrow Y \\ def \ \Gamma \ in \\ \\ let \ assocx : \forall Z_1 [\exists Z_2. \begin{pmatrix} assoc \leq Z_2 \rightarrow Z_1 \\ x \leq Z_2 \end{pmatrix}]. \ Z_1 \ in \\ \\ \exists Y_1 Y_2. \begin{pmatrix} Y = Y_1 \times Y_2 \\ \forall i \quad \exists Z_2. (assocx \leq Z_2 \rightarrow Y_i \land I_i \leq Z_2) \end{pmatrix}$$

(The index *i* ranges over $\{1,2\}$.)

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplific	ation				

Constraint solving can be viewed as a *rewriting process* that exploits *equivalence laws*. Because equivalence is, by construction, a *congruence*, rewriting is permitted within an arbitrary context.

For instance, environment access is allowed by the law

 $let x : \sigma in C[x \leq T] \equiv let x : \sigma in C[\sigma \leq T]$

where C is a context that does not bind x.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplific	ation, con	tinued			

Thus, within the context def Γ_0 ; Γ in [], the constraint:

$$\begin{array}{c} \text{assoc} \leq Z_2 \rightarrow Z_1 \\ \times \leq Z_2 \end{array} \right)$$

is equivalent to:

$$\begin{pmatrix} \exists XY.(X \to \text{list}(X \times Y) \to Y = Z_2 \to Z_1) \\ X_0 = Z_2 \end{pmatrix}$$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplific	ation, con	tinued			

By first-order unification, the constraint:

$$\exists Z_2. (\exists XY. (X \rightarrow \text{list} (X \times Y) \rightarrow Y = Z_2 \rightarrow Z_1) \land X_0 = Z_2)$$

simplifies down successively to:

$$\exists Z_2. (\exists XY. (X = Z_2 \land \text{list} (X \times Y) \rightarrow Y = Z_1) \land X_0 = Z_2)$$
$$\exists Z_2. (\exists Y. (\text{list} (Z_2 \times Y) \rightarrow Y = Z_1) \land X_0 = Z_2)$$
$$\exists Y. (\text{list} (X_0 \times Y) \rightarrow Y = Z_1)$$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplific	ation, con	tinued			

The constrained type scheme:

 $\forall Z_1[\exists Z_2.(assoc \leq Z_2 \rightarrow Z_1 \land x \leq Z_2)]. Z_1$

is thus equivalent to:

$$\forall Z_1[\exists Y. (list (X_0 \times Y) \rightarrow Y = Z_1)]. Z_1$$

which can also be written:

$$\forall Z_1 \Upsilon [list (X_0 \times \Upsilon) \to \Upsilon = Z_1]. Z_1$$

$$\forall \Upsilon. list (X_0 \times \Upsilon) \to \Upsilon$$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplifica	ation, con	tinued			

The initial constraint has now been simplified down to:

The simplification work spent on assocx's type scheme was well worth the trouble, because we are now going to *duplicate* the simplified type scheme.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplific	ation, cont	tinued			

The sub-constraint:

 $\exists Z_2. (assocx \leq Z_2 \rightarrow Y_i \land I_i \leq Z_2)$

where $i \in \{1, 2\}$, is rewritten:

 $\exists Z_2. (\exists Y. (list (X_0 \times Y) \to Y = Z_2 \to Y_i) \land X_i = Z_2)$ $\exists Y. (list (X_0 \times Y) \to Y = X_i \to Y_i)$ $\exists Y. (list (X_0 \times Y) = X_i \land Y = Y_i)$ $list (X_0 \times Y_i) = X_i$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplifica	ation, con	tinued			

The initial constraint has now been simplified down to:

Now, the context def Γ in let assocx :... in [] can be dropped, because the constraint that it applies to contains no occurrences of x, l_1 , l_2 , or assocx.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplific	ation, con	tinued			

The constraint becomes:

$$\exists X_0 X_1 X_2 Y. \left(\begin{array}{c} X = X_0 \to X_1 \to X_2 \to Y \\ \exists Y_1 Y_2. \left(\begin{array}{c} Y = Y_1 \times Y_2 \\ \forall i \quad \text{list} (X_0 \times Y_i) = X_i \end{array}\right) \end{array}\right)$$

that is:

$$\exists X_0 X_1 X_2 Y Y_1 Y_2. \begin{pmatrix} X = X_0 \rightarrow X_1 \rightarrow X_2 \rightarrow Y \\ Y = Y_1 \times Y_2 \\ \forall i \quad \text{list} (X_0 \times Y_i) = X_i \end{pmatrix}$$

and, by eliminating a few auxiliary variables:

 $\exists X_0 Y_1 Y_2. \left(X = X_0 \rightarrow \text{list} \left(X_0 \times Y_1 \right) \rightarrow \text{list} \left(X_0 \times Y_2 \right) \rightarrow Y_1 \times Y_2 \right)$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Simplific	ation, the	end			

We have shown the following equivalence between constraints:

$$def \Gamma_0 \text{ in } \llbracket a : X \rrbracket$$

$$\equiv \exists X_0 Y_1 Y_2. (X = X_0 \rightarrow \text{list} (X_0 \times Y_1) \rightarrow \text{list} (X_0 \times Y_2) \rightarrow Y_1 \times Y_2)$$

That is, the principal type scheme of a relative to $\Gamma_{\rm O}$ is

$$\begin{aligned} \|a\| &= \forall X [[[a:X]]]. X \\ &= \forall X_0 Y_1 Y_2. X_0 \rightarrow \text{list} (X_0 \times Y_1) \rightarrow \text{list} (X_0 \times Y_2) \rightarrow Y_1 \times Y_2 \end{aligned}$$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Rewriting	strategies	5			

Again, constraint solving can be explained in terms of a *small-step rewrite system*. Again, one checks that every step is meaning-preserving, that the system is normalizing, and that every normal form is either literally "false" or satisfiable.

Different constraint solving *strategies* lead to different behaviors in terms of complexity, error explanation, etc.

See ATTAPL for details on constraint solving [Pottier and Rémy, 2005]. See Jones [1999] for a different presentation of type inference, in the context of Haskell.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Rewriting	strategie	6			

In all reasonable strategies, the left-hand side of a let constraint is simplified *before* the let form is expanded away.

This corresponds, in Algorithm \mathcal{J} , to computing a principal type scheme before examining the right-hand side of a let construct.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Complexit	ty				

Type inference for ML is DEXPTIME-complete [Kfoury et al., 1990; Mairson, 1990], so any constraint solver has exponential complexity.

Nevertheless, under the hypotheses that types have bounded size and let forms have bounded left-nesting depth, constraints can be solved in linear time [McAllester, 2003].

This explains why ML type inference works well in practice.

Introduction Simple types Core ML Type annotations Recursive Types System F An alternative presentation of constraint generation

Using principal contrained type schemes and the following equivalence $(a) \stackrel{\Delta}{=} \forall X [[[a:X]]] . X \qquad [[a:T]] \equiv (a) \leq T$

we call also present contraints as follows:

$$\begin{aligned} (|x|) &= \forall X[x \leq X]. X \\ (|\lambda x. a|) &= \forall X_1 X_2 [def x : X_1 in (|a|) \leq X_2]. X_1 \rightarrow X_2 \\ & \text{if } X_1, X_2 \# a \end{aligned}$$
$$\begin{aligned} (|a_1 a_2|) &= \forall X_1 X_2 [(|a_1|) \leq X_1 \rightarrow X_2 \land (|a_2|) \leq X_1]. X_2 \\ & \text{if } X_1, X_2 \# a_1, a_2 \end{aligned}$$
$$(|et x = a_1 in a_2|) &= \forall X [let x : (|a_1|) in (|a_2|) \leq X]. X$$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
From	implicitly-type	ed to	explicitly-typed	terms	

Every node of the source program is thus mapped to a principal type scheme relative to its program context.

By default, constraint resolution removes solved contains.

However, solved constraints may be kept: principal constrained type schemes become principal type schemes which points back to source program nodes and may used to build a decorated source term.

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Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
On type	annotatic	ns			

Damas and Milner's type system has *principal types:* at least in the core language, no type information is required.

This is very lightweight, but a bit extreme: sometimes, it is useful to write types down, and use them as *machine-checked documentation*.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Syntax	for type	annotations	5		

Let us, then, allow programmers to annotate a term with a type: $a ::= \dots | (a:T)$

Typing and constraint generation are obvious:

Annot

$$\frac{\Gamma \vdash a:T}{\Gamma \vdash (a:T):T} \qquad [[(a:T):T']] = [[a:T]] \land T = T'$$

Type annotations are *erased* prior to runtime, so the operational semantics is not affected.

(Erasure of type annotations preserves well-typedness.)

The constraint [(a:T):T'] implies the constraint [a:T'].

That is, in terms of type inference, type annotations are restrictive: they lead to a principal type that is less general, and possibly even to ill-typedness.

For instance, $\lambda x. x$ has principal type scheme $\forall X. X \rightarrow X$, whereas $(\lambda x. x: int \rightarrow int)$ has principal type scheme int \rightarrow int, and $(\lambda x. x: int \rightarrow bool)$ is ill-typed.

IntroductionSimple typesCore MLType annotationsRecursive TypesSystem FType variables within type annotations?

Does it make sense for a type annotation to contain a type variable, as in, say:

$$\begin{aligned} & (\lambda x. x : X \to X) \\ & (\lambda x. x + 1 : X \to X) \\ & \text{let } f = (\lambda x. x : X \to X) \text{ in } (f \ O, f \ true) \end{aligned}$$

If so, what does it mean?

Short answer: it does not mean anything, because X is unbound. "There is no such thing as a free variable" (Alan Perlis).

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
How and	d where				

Does it make sense for a type annotation to contain a type variable, as in, say:

$$\begin{aligned} & (\lambda x. x : X \to X) \\ & (\lambda x. x + 1 : X \to X) \\ & \text{let } f = (\lambda x. x : X \to X) \text{ in } (f \ O, f \ \text{true}) \end{aligned}$$

If so, what does it mean?

A longer answer:

It is necessary to specify how and where type variables are bound.

 Introduction
 Simple types
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 Type
 variables
 within
 type
 annotations?
 Support

Does it make sense for a type annotation to contain a type variable, as in, say:

$$\begin{aligned} & (\lambda x. x : X \to X) \\ & (\lambda x. x + 1 : X \to X) \\ & \text{let } f = (\lambda x. x : X \to X) \text{ in } (f \ O, f \ \text{true}) \end{aligned}$$

If so, what does it mean?

How is X bound?

If X is existentially bound, or flexible, then both $(\lambda x. x : X \to X)$ and $(\lambda x. x + 1 : X \to X)$ should be well-typed.

If it is universally bound, or rigid, only the former should be well-typed.

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Where					

Does it make sense for a type annotation to contain a type variable, as in, say:

$$\begin{aligned} & (\lambda x. x : X \to X) \\ & (\lambda x. x + 1 : X \to X) \\ & \text{let } f = (\lambda x. x : X \to X) \text{ in } (f \ O, f \ \text{true}) \end{aligned}$$

If so, what does it mean?

Where is X bound?

If X is bound within the left-hand side of this "let" construct, then this code:

let
$$f = (\lambda x. x : X \rightarrow X)$$
 in $(f O, f true)$

should be well-typed.

On the other hand, if X is bound *outside* this "let" form, then this code should be ill-typed, since no *single* ground value of X is suitable.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Binding	type varia	ibles			

Let's allow programmers to *explicitly bind* type variables:

It now makes sense for a type annotation (a:T) to contain free type variables.

Terms a can now contain free type variables, so some side conditions have to be updated (e.g., $\bar{X} \# \Gamma$, a in Gen).

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Binding	type varia	bles			

The typing rules are as follows: $\frac{\Gamma \vdash [\vec{X} \mapsto \vec{T}]a:T}{\Gamma \vdash \exists \bar{X}.a:T} \qquad \frac{\Gamma \vdash a:T}{\Gamma \vdash \forall \bar{X}.a:\forall \bar{X}.T} \qquad \left(\begin{array}{c} Gen \\ \Gamma \vdash a:T \quad \bar{X} \# \Gamma, a \\ \hline \Gamma \vdash a:\forall \bar{X}.T \end{array}\right)$

Again, these constructs are erased prior to runtime.

(Why is this sound? Easy exercise: define the erasure of a term, and prove that the erasure of a well-typed term is well-typed.)

IntroductionSimple typesCore MLType annotationsRecursive TypesSystem FConstraint generation:existential case

Constraint generation for the existential form is straightforward:

$$\llbracket (\exists \bar{X}.a):T \rrbracket = \exists \bar{X}.\llbracket a:T \rrbracket \text{ if } \bar{X} \# T$$

The type annotations inside a contain free occurrences of \bar{X} . Thus, the constraint $[\![a:T]\!]$ contains such occurrences as well. They are bound by the existential quantifier.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Constrain	t generatio	on: ex	istential case		

For instance, the expression:

$$\lambda x_1. \lambda x_2. \exists X. ((x_1 : X), (x_2 : X))$$

has principal type scheme $\forall X. X \rightarrow X \rightarrow X \times X$. Indeed, the generated constraint contains the pattern:

$$\exists X.(\llbracket x_1 : X \rrbracket \land \llbracket x_2 : X \rrbracket \land \ldots)$$

which requires x_1 and x_2 to share a common (unspecified) type.

A term *a* has type scheme, say, $\forall X. X \rightarrow X$ if and only if *a* has type $X \rightarrow X$ for every instance of X, or, equivalently, for an abstract X.

To express this in terms of constraints, we introduce *universal quantification* in the constraint language:

 $C ::= \dots | \forall X.C$

Its interpretation is standard.

(To solve these constraints, we will use an extension of the unification algorithm called unification under a mixed prefix—see \bigcirc forward.)

The need for universal quantification in constraints arises when polymorphism is *required* by the programmer, as opposed to *inferred* by the system.

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Constrair	it generat	tion: uni	versal case		

Constraint generation for the universal form is somewhat subtle. A naive definition *fails* (why?):

$$\llbracket \forall \bar{X}.a:T \rrbracket = \forall \bar{X}.\llbracket a:T \rrbracket \quad \text{if } \bar{X} \# T \qquad (Wrong)$$

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 Constraint generation:
 universal case

Constraint generation for the universal form is somewhat subtle. A naive definition *fails*:

$$\llbracket \forall \bar{X}.a:T \rrbracket = \forall \bar{X}.\llbracket a:T \rrbracket \qquad \text{if } \bar{X} \# T \qquad (Wrong)$$

This requires T to be simultaneously equal to all of the types that a assumes when \bar{X} varies.

For instance, with this incorrect definition, one would have:

$$\begin{bmatrix} \forall X. (\lambda x. x : X \to X) : int \to int \end{bmatrix} = \forall X. \begin{bmatrix} (\lambda x. x : X \to X) : int \to int \end{bmatrix}$$
$$\equiv \forall X. (\begin{bmatrix} \lambda x. x : X \to X \end{bmatrix} \land X = int)$$
$$\equiv \forall X. (true \land X = int)$$
$$\equiv false$$

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Constrai	nt generat	tion: univ	versal case		

A correct definition is:

This requires a to be well-typed for all instances of \bar{X} and requires T to be a valid type for a under some instance of \bar{X} .

A problem with this definition is...

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Constraint	t generatio	on: unive	ersal case		

A correct definition is:

$$[\![\forall \bar{X}.a:T]\!] = \forall \bar{X}.\exists Z.[\![a:Z]\!] \land \exists \bar{X}.[\![a:T]\!]$$

This requires a to be well-typed for all instances of \bar{X} and requires T to be a valid type for a under some instance of \bar{X} .

A problem with this definition is...

The term *a* is duplicated! This can lead to exponential complexity. Fortunately, this can be avoided modulo a slight extension of the constraint language [Pottier and Rémy, 2003, p. 112].

The solution defines:

$$\left[\!\left[\forall \bar{X}.a: \mathcal{T}\right]\!\right] = \operatorname{let} x: \forall \bar{X}, Y\left[\!\left[\!\left[a:Y\right]\!\right]\!\right]\!. Y \text{ in } x \leq \mathcal{T}$$

where the new constrain form satisfies the equivalence:

 $\mathsf{let}\, \mathsf{x}:\,\forall\vec{X},\vec{Y}[\mathcal{C}_1].\,\mathcal{T}\,\,\mathsf{in}\,\,\mathcal{C}2\equiv\,\forall\vec{X}.\,\exists\vec{Y}.\,\mathcal{C}_1\wedge\mathsf{def}\,\,\mathsf{x}:\,\forall\vec{X},\vec{Y}[\mathcal{C}_1].\,\mathcal{T}\,\,\mathsf{in}\,\,\mathcal{C}2$

Introduction Simple types Core ML Type annotations Recursive Types System F Type schemes as annotations

Annotating a term with a type scheme, rather than just a type, is now just syntactic sugar:

$$(a: \forall \overline{X}.T)$$
 stands for $\forall \overline{X}.(a:T)$ if $\overline{X} \# a$

In that particular case, constraint generation is in fact simpler:

$$\llbracket (a:\forall \bar{X}.T):T' \rrbracket \equiv \forall \bar{X}.\llbracket a:T \rrbracket \land (\forall \bar{X}.T) \preceq T'$$

(Exercise: check this equivalence.)

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Examples					

A correct example:

$$\begin{bmatrix} (\exists X.(\lambda x. x + 1 : X \to X)) : int \to int \end{bmatrix}$$

= $\exists X. \begin{bmatrix} (\lambda x. x + 1 : X \to X) : int \to int \end{bmatrix}$
= $\exists X. (X = int)$
= true

The system infers that X must be int. Because X is a local type variable, it does not appear in the final constraint.

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Examples					

An incorrect example:

$$\begin{bmatrix} (\forall X.(\lambda x. x + 1 : X \to X)) : int \to int \end{bmatrix}$$

$$\vdash \forall X.\exists Z.\llbracket(\lambda x. x + 1 : X \to X) : Z \rrbracket$$

$$\equiv \forall X.\exists Z.(X = int \land X \to X = Z)$$

$$\equiv \forall X.X = int$$

$$\equiv false$$

The system *checks* that X is used in an abstract way, which is not the case here, since the code implicitly assumes that X is int.

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Examples					

A correct example:

 $\begin{bmatrix} (\forall X.(\lambda x. x : X \to X)) : int \to int \end{bmatrix}$ = $\forall X.\exists Z. \begin{bmatrix} (\lambda x. x : X \to X) : Z \end{bmatrix} \land \exists X. \begin{bmatrix} (\lambda x. x : X \to X) : int \to int \end{bmatrix}$ = $\forall X.\exists Z. X \to X = Z \land \exists X. X = int$

The system *checks* that X is used in an abstract way, which is indeed the case here.

It also checks that, if X is appropriately instantiated, the code admits the expected type int \rightarrow int.

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Examples					

An incorrect example:

 $\begin{array}{l} \left[\exists X.(\operatorname{let} f = (\lambda x. x : X \to X) \text{ in } (f O, f \operatorname{true})) : Z \right] \\ \equiv & \exists X.(\operatorname{let} f : X \to X \text{ in } \exists Z_1 Z_2.(f \le \operatorname{int} \to Z_1 \land f \le \operatorname{bool} \to Z_2 \land Z_1 \times Z_2 = Z)) \\ \equiv & \exists X Z_1 Z_2.(X \to X = \operatorname{int} \to Z_1 \land X \to X = \operatorname{bool} \to Z_2 \land Z_1 \times Z_2 = Z) \\ \Vdash & \exists X.(X = \operatorname{int} \land X = \operatorname{bool}) \\ \equiv & \operatorname{false} \end{array}$

X is bound outside the let construct; f receives the monotype $X \rightarrow X$.

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Examples					

A correct example:

 $\begin{bmatrix} \operatorname{let} f = \exists X. (\lambda x. x : X \to X) \text{ in } (f O, f \operatorname{true}) : Z \end{bmatrix}$ $\equiv \operatorname{let} f : \forall Y [\exists X. (X \to X = Y)]. Y \text{ in}$ $\exists Z_1 Z_2. (f \leq \operatorname{int} \to Z_1 \land f \leq \operatorname{bool} \to Z_2 \land Z_1 \times Z_2 = Z)$ $\equiv \operatorname{let} f : \forall X. X \to X \text{ in}$ $\exists Z_1 Z_2. (\ldots)$ $\equiv \exists Z_1 Z_2. (\operatorname{int} = Z_1 \land \operatorname{bool} = Z_2 \land Z_1 \times Z_2 = Z)$ $\equiv \operatorname{int} \times \operatorname{bool} = Z$

X is bound within the let construct; the term $\exists X.(\lambda x. x : X \to X)$ has the same principal type scheme as $\lambda x. x$, namely $\forall X. X \to X$; f receives the type scheme $\forall X. X \to X$. Introduction Simple types Core ML Type annotations Recursive Types System F

Type annotations in the real world

For historical reasons, in Objective Caml, type variables are not explicitly bound. (Retrospectively, that's *bad!*) They are implicitly *existentially* bound at the nearest enclosing toplevel let construct.

In Standard ML, type variables are implicitly *universally* bound at the nearest enclosing toplevel let construct.

In Glasgow Haskell, type variables are implicitly existentially bound within patterns: 'A pattern type signature brings into scope any type variables free in the signature that are not already in scope' [Peyton Jones and Shields, 2004].

Constraints help understand these varied design choices uniformly.

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Monomor	phic recu	rsion			

Recall the typing rule for recursive functions:

 $\frac{\Gamma, f: T \vdash \lambda x. a: T}{\Gamma \vdash \mu f. \lambda x. a: T}$

It leads to the following derived typing rule:

LetRec $\Gamma, f: T_1 \vdash \lambda x. a_1 : T_1 \qquad \bar{X} \# \Gamma, a_1$ $\Gamma, f: \forall \bar{X}. T_1 \vdash a_2 : T_2$

 $\Gamma \vdash \text{let rec } f \times = a_1 \text{ in } a_2 : T_2$

Any comments?

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Monomor	rphic recu	rsion			

These rules require occurrences of f to have monomorphic type within the recursive definition (that is, within $\lambda x. a_1$).

This is visible also in terms of type inference. The constraint

$$\llbracket \text{let rec } f \times = a_1 \text{ in } a_2 : T \rrbracket$$

is equivalent to

let $f: \forall XY [let f: X \rightarrow Y; x: X in [a_1:Y]] . X \rightarrow Y in [a_2:T]$

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Monomo	rphic recu	rsion			

This is problematic in some situations, most particularly when defining functions over *nested algebraic data types* [Bird and Meertens, 1998; Okasaki, 1999].

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Polymorp	hic recure	ion			

This problem is solved by introducing *polymorphic recursion*, that is, by allowing μ -bound variables to receive a polymorphic type scheme:

FixAbsPoly	LetRecPoly	
Γ,f : S ⊢ λx.a : S	Γ,f : S ⊢ λx.a ₁ : S	Γ,f: S⊢ a2:T
Γ ⊢ µf.λx.a : S	$\Gamma \vdash \text{let rec } f \times f$	= a ₁ in a ₂ : T

This extension of ML is due to Mycroft [1984].

In System F, there is no problem to begin with; no extension is necessary.

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Polymorp	hic recure	sion			

Polymorphic recursion alters, to some extent, Damas and Milner's type system.

Now, not only *let-bound*, but also μ -bound variables receive type schemes. The type system is no longer equivalent, up to reduction to let-normal form, to simply-typed λ -calculus.

This has two consequences:

• *monomorphization*, a technique employed in some ML compilers [Tolmach and Oliva, 1998; Cejtin et al., 2007], is no longer possible;

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Polymorp	phic recure	sion			

Polymorphic recursion alters, to some extent, Damas and Milner's type system.

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This has two consequences:

- *monomorphization*, a technique employed in some ML compilers [Tolmach and Oliva, 1998; Cejtin et al., 2007], is no longer possible;
- type inference becomes problematic!

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Polymorp	phic recure	sion			

Type inference for ML with polymorphic recursion is undecidable [Henglein, 1993]. It is equivalent to the undecidable problem of *semi-unification*.

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Polymorp	phic recure	sion			

Yet, type inference in the presence of polymorphic recursion can be made simple. (How?)

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Polymorp	phic recure	sion			

Yet, type inference in the presence of polymorphic recursion can be made simple. (How?)

By relying on a mandatory type annotation. The rules become:

FixAbsPolyLetRecPoly $\Gamma, f: S \vdash \lambda x. a: S$ $\Gamma, f: S \vdash \lambda x. a_1 : S$ $\Gamma, f: S \vdash a_2 : T$ $\Gamma \vdash \mu(f: S).\lambda x. a: S$ $\Gamma \vdash \text{let rec } (f: S) = \lambda x. a_1 \text{ in } a_2 : T$

The type scheme S no longer has to be guessed.

With this feature, contrary to what was said earlier (back), type annotations are not just restrictive: they are sometimes required for type inference to succeed.

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Polymor	ohic recure	aion			

The constraint generation rule becomes:

 $\llbracket \text{let rec } (f:S) = \lambda x. a_1 \text{ in } a_2:T \rrbracket = ?$

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Polymorp	hic recure	sion			

The constraint generation rule becomes:

$$\llbracket \text{let rec } (f:S) = \lambda x. a_1 \text{ in } a_2 : T \rrbracket = \text{let } f:S \text{ in } (\llbracket \lambda x. a_1 : S \rrbracket \land \llbracket a_2 : T \rrbracket)$$

It is clear that f receives type scheme S both *inside and outside* of the recursive definition.

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Unificatior	ı under	а	mixed	prefix		

Unification under a mixed prefix means unification in the presence of both existential and universal quantifiers.

We extend the basic unification algorithm with support for universal quantification.

The solved forms are unchanged: universal quantifiers are always *eliminated.*

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Unificati	on under a	mixed	prefix		

In short, in order to reduce $\forall \bar{X}.C$ to a solved form, where C is itself a solved form:

- if a rigid variable is equated with a constructed type, fail;
- if two rigid variables are equated, fail;
- if a free variable dominates a rigid variable, fail;
- otherwise, one can decompose C as $\exists \overline{Y}.(C_1 \land C_2)$, where $\overline{X}\overline{Y} \# C_1$ and $\exists \overline{Y}.C_2 \equiv \text{true}$; in that case, $\forall \overline{X}.C$ reduces to just C_1 .

See [Pottier and Rémy, 2003, p. 109] for details.



Here are examples of the situations described on the previous slide:

- $\forall X.\exists YZ.(X = Y \rightarrow Z)$ is false;
- $\forall XY.(X = Y)$ is false;
- $\forall X.\exists Y.(Z = X \rightarrow Y)$ is false;
- $\forall X.\exists YZ_1Z_2.(Y = X \rightarrow Z \land Z = Z_1 \rightarrow Z_2)$ reduces to just $\exists Z_1Z_2.(Z = Z_1 \rightarrow Z_2)$. The constraint $\forall X.\exists Y.(Y = X \rightarrow Z)$ is equivalent to true.

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Examples					

Objective Caml implements a form of unification under a mixed prefix:

```
bash$ ocaml
# let module M : sig val id : 'a \rightarrow 'a end
= struct let id x = x + 1 end
in M.id;;
Values do not match: val id : int \rightarrow int
is not included in val id : 'a \rightarrow 'a
```

This example gives rise to a constraint of the form $\forall X.X = int.$

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Examples					

Here is another example:

```
bash$ ocaml

# let r = ref (fun \times \rightarrow \times) in

let module M : sig val id : 'a \rightarrow 'a end

= struct let id = !r end

in M.id;;

Values do not match: val id : '_a \rightarrow '_a

is not included in val id : 'a \rightarrow 'a
```

This example gives rise to a constraint of the form $\exists Y.\forall X.X = Y$.

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Recursive	types				

Product and sum types alone do not allow describing *data structures* of *unbounded size*, such as lists and trees.

Indeed, if the grammar of types is $T ::= \text{unit} | T \times T | T + T$, then it is clear that every type describes a *finite* set of values.

For every k, the type of lists of length at most k is expressible using this grammar. However, the type of lists of unbounded length is not.

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The following definition is inherently *recursive*:

"A list is either empty or a pair of an element and a list." We need something like this: list $X \Leftrightarrow$ unit + $X \times$ list XBut what does \diamond stand for? Is it equality, or some kind of isomorphism? There are two standard approaches to recursive types, dubbed the equi-recursive and iso-recursive approaches.

In the equi-recursive approach, a recursive type is equal to its unfolding.

In the iso-recursive approach, a recursive type and its unfolding are related via explicit *coercions*.

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Equi-rec	ursive type	85			

In the equi-recursive approach, the usual syntax of types:

 $T ::= X \mid F \vec{T}$

is no longer interpreted inductively. Instead, types are the *regular trees* built on top of this signature.

Introduction Simple types Core ML Type annotations Recursive Types System F Finite syntax for equi-recursive types

If desired, it is possible to use finite syntax for recursive types: $T ::= X \mid \mu X. (F \vec{T})$

We do not allow the seemingly more general $\mu X.T$, because $\mu X.X$ is meaningless, and $\mu X.Y$ or $\mu X.\mu Y.T$ are useless. If we write $\mu X.T$, it should be understood that T is *contractive*, that is, T is a type constructor application.

For instance, the type of lists of elements of type X is:

 μ Y.(unit + X × Y)

Introduction Simple types Core ML Type annotations Recursive Types System F Finite syntax for equi-recursive types

Each type in this syntax denotes a unique regular tree, sometimes known as its *infinite unfolding*. Conversely, every regular tree can be expressed in this notation (possibly in more than one way).

If one builds a type-checker on top of this finite syntax, then one must be able to *decide* whether two types are *equal*, that is, have identical infinite unfoldings.

This can be done efficiently, either via the algorithm for comparing two DFAs, or by unification. (The latter approach is simpler, faster, and extends to the type inference problem.)

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Finite	syntax for	equi-recu	rsive types		

One can also prove [Brandt and Henglein, 1998] that equality is the least congruence generated by the following two rules:

Fold/Unfold

$$\mu X.T = [X \mapsto \mu X.T]T$$
Uniqueness
$$T_1 = [X \mapsto T_1]T \qquad T_2 = [X \mapsto T_2]T$$

$$T_1 = T_2$$

In both rules, T must be contractive.

This axiomatization does not directly lead to an efficient algorithm for deciding equality, though.

Introduction Simple types Core ML Type annotations Recursive Types System F Type soundness for equi-recursive types

In the presence of equi-recursive types, structural induction on types is no longer permitted — but *we never used it* anyway.

It remains true that $F \vec{T}_1 = F \vec{T}_2$ implies $\vec{T}_1 = \vec{T}_2$ — this was used in our Subject Reduction proofs.

It remains true that $F_1 \vec{T}_1 = F_2 \vec{T}_2$ implies $F_1 = F_2$ — this was used in our Progress proofs.

So, the reasoning that leads to type soundness is unaffected.

(Exercise: prove type soundness for the simply-typed λ -calculus in Coq. Then, change the syntax of types from Inductive to CoInductive.)

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How is type inference adapted for equi-recursive types?

The syntax of constraints is unchanged: they remain systems of equations between finite first-order types, without μ 's. Their *interpretation* changes: they are now interpreted in a universe of regular trees.

As a result,

- constraint generation is unchanged;
- constraint solving is adapted by removing the occurs check.

(Exercise: describe solved forms and show that every solved form is either false or satisfiable.)

Introduction Simple types Core ML Type annotations Recursive Types System F Type inference for equi-recursive types Here is a function that measures the length of a list: $\mu length.\lambda xs.case xs of$ $\lambda().O$ $[] \lambda(x, xs).1 + length xs$

Type inference gives rise to the cyclic equation:

 $Y = unit + X \times Y$

where length has type $Y \rightarrow int$.

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 Type inference for equi-recursive types

That is, length has principal type scheme:

 $\forall X. (\mu Y. unit + X \times Y) \rightarrow int$

or, equivalently, principal constrained type scheme:

 $\forall X [Y = unit + X \times Y]. Y \rightarrow int$

The cyclic equation that characterizes lists was never provided by the programmer, but was inferred.

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Objective Caml implements equi-recursive types upon explicit request:

```
bash$ ocaml -rectypes
# type ('a, 'b) sum = Left of 'a | Right of 'b;;
type ('a, 'b) sum = Left of 'a | Right of 'b
# let rec length xs =
match xs with
| Left () \rightarrow O
| Right (x, xs) \rightarrow 1 + length xs ;;
val length : ((unit, 'b * 'a) sum as 'a) \rightarrow int = (fun)
```

Quiz: why is -rectypes only an option?

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Drawbacks of equi-recursive types

Equi-recursive types are simple and powerful. In practice, however, they are perhaps *too expressive:*

```
bash$ ocaml -rectypes
# let rec map f = function
  | [1 \rightarrow [1]
  | x :: xs \rightarrow map f x :: map f xs;;
val map : 'a \rightarrow ('b list as 'b) \rightarrow ('c list as 'c) = (fun)
# map (fun x \rightarrow x + 1) [ 1; 2 ];;
This expression has type int but is used with type 'a list as 'a
# map () [[];[[]]];;
-: 'a list as 'a = [[]; [[]]]
```

Equi-recursive types allow this nonsensical version of map to be accepted, thus delaying the detection of a programmer error.

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Half a	pint of equ	ui-recursi	ve types		

```
Quiz: why is this accepted?

bash$ ocaml

# let f \times = x#hello x;;

val f : (< hello : 'a \rightarrow 'b; ... > as 'a) \rightarrow 'b = (fun)
```

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lso-recur	rsive types	5			

In the iso-recursive approach, the user is allowed to introduce new type constructors D via (possibly mutually recursive) declarations:

$$D\vec{X} \approx T$$
 (where $ftv(T) \subseteq \bar{X}$)

Each such declaration adds a unary constructor fold_D and a unary destructor unfold_D with the following types:

fold _D	:	$\forall \bar{X}.T \to D \bar{X}$
unfold _D	:	$\forall \bar{X}. D \vec{X} \to T$

and the reduction rule:

 $\mathsf{unfold}_{D} (\mathsf{fold}_{D} w) \longrightarrow w$

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lso-recu	rsive types	5			

Ideally, iso-recursive types should not have any runtime cost.

One solution is to compile constructors and destructors away into a target language with equi-recursive types.

Another solution is to see iso-recursive types as a restriction of equi-recursive types where the source language does not have equi-recursive types but instead two unary destructors fold_D and unfold_D with the semantics of the identity function.

Subject reduction does not hold in the source language, but only in the full language with iso-recursive types. Applications of destructors can also be reduced at compile time.

Note that iso-recursive types are less expressive than equi-recursive types, as there is no counter-part to the Uniqueness typing rule.

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lso-recu	rsive lists				

A parametrized, iso-recursive type of lists is:

list $X \approx \text{unit} + X \times \text{list} X$

The empty list is:

 $fold_{list}$ (inj₁ ()) : $\forall X. list X$

A function that measures the length of a list is:

$$\left(\begin{array}{c} \mu length.\lambda xs.case (unfold_{list} xs) of \\ \lambda().O \\ [] \lambda(x,xs).1 + length xs \end{array}\right) : \forall X. list X \to int$$

One folds upon construction and unfolds upon deconstruction.

In the iso-recursive approach, types remain finite. The type list X is just an application of a type constructor to a type variable.

As a result, type inference is unaffected. The occurs check remains.

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Algebraic	data 1	types			

Algebraic data types result of the fusion of iso-recursive types with structural, labeled products and sums.

This suppresses the *verbosity* of explicit folds and unfolds as well as the *fragility* and inconvenience of numeric indices — instead, named *record fields* and *data constructors* are used.

For instance,

 $fold_{list}(inj_{1}())$ is replaced with Nil()

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An algebraic data type constructor D is introduced via a record type or variant type definition:

$$D\vec{X} \approx \prod_{\ell \in L} \ell : T_{\ell}$$
 or $D\vec{X} \approx \sum_{\ell \in L} \ell : T_{\ell}$

L denotes a finite set of record labels or data constructors.

Algebraic data type definitions can be mutually recursive.

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The record type definition $D\vec{X} \approx \prod_{\ell \in L} \ell : T_{\ell}$ introduces syntax for constructing and deconstructing records:

$$C ::= \dots \mid \{\ell = \cdot\}_{\ell \in L} \qquad \qquad d ::= \dots \mid \cdot.\ell$$

With the following types

$$\{ \ell_1 = \cdot, \dots, \ell_n \} : \quad \forall \vec{X} . \ T_{\ell_1} \to \dots T_{\ell_n} \to D \ \vec{X} \\ \cdot . \ell : \quad \forall \vec{X} . D \ \vec{X} \to T_{\ell}$$

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 Core ML
 Type annotations
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 System F

 Effects of a variant type declaration

The variant type definition $D\vec{X} \approx \sum_{\ell \in L} \ell : T_{\ell}$ introduces syntax for constructing and deconstructing variants:

 $C ::= \dots \mid \ell \qquad \qquad d ::= \dots \mid case \cdot of \ [\ell : \cdot]_{\ell \in L}$

With the following types:

case
$$\cdot$$
 of $[\ell_1 : \cdot [] \dots \ell_n : \cdot] : \forall \vec{X} Y. D \vec{X} \to (T_{\ell_1} \to Y) \to \dots (T_{\ell_n} \to Y) \to Y$
 $\cdot \ell : \forall \vec{X}. T_\ell \to D \vec{X}$

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An exan	nple: lists				
Here is a	n algebraic d	ata type o	f lists:		

```
list X \approx Nil : unit + Cons : X \times list X
```

```
This gives rise to:
```

```
\begin{array}{ll} case \, \cdot \, of \, [Nil: \cdot \, [] \, \dots Cons: \cdot] : & \forall XY. \, \text{list } X \to (\text{unit} \to Y) \to \\ & & ((X \times \text{list } X) \to Y) \to Y \\ Nil: & \forall X. \, \text{unit} \to \text{list } X \\ Cons: & \forall X. \, (X \times \text{list } X) \to \text{list } X \end{array}
```

A function that measures the length of a list is:

$$\left(\begin{array}{c} \mu \text{length.} \lambda \text{xs.case xs of} \\ \text{Nil}: \lambda().O \\ \square \text{ Cons}: \lambda(\text{x,xs}).1 + \text{length xs} \end{array}\right): \forall X. \text{ list } X \to \text{int}$$

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A word	on mutable	fields			

In Objective Caml, a record field can be marked *mutable*. This introduces an extra binary destructor for writing this field:

$$(\cdot, \ell \leftarrow \cdot): \forall \vec{X}. D \vec{T} \rightarrow T_{\ell} \rightarrow \text{unit}$$

This also makes record construction a destructor since, when fully applied it is *not a value* but it allocates a piece of store and returns its location.

Thus, due to the value restriction, the type of such expressions cannot be generalized.

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- Type inference for simply-typed λ -calculus
- Type inference for ML
 - Constraint-based type inference for ML
 - Constraint solving by example
- Type annotations
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 - Unification under a mixed prefix
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Full type	inference				

Type inference has long been an open problem for System F, until Wells [1999] showed that it is in fact undecidable by showing it is equivalent to the semi-unification problem which was earlier proved undecidable.

Type-checking in explicitly-typed System F is indeed feasible and easy (still, an implementation must be careful with renaming of variables when applying substitutions).

However, we have seen that programming with fully-explicit types is unpractical.

Several solutions for *partial type inference* are used in practice. They may alleviate the need for a lot of redundant type annotations. However, none of them is fully satisfactory. Introduction Simple types Core ML Type annotations Recursive Types System F Type inference and second-order unification

The full type-inference problem is not directly related to second-order unification but rather to semi-unification.

However, it becomes equivalent to second-order unification if the positions of type abstractions and type applications are explicit. That is, terms are

$$t ::= x | \lambda x : ?.t | tt | \land ?.t | t?$$

where the question marks stand for type variables names and types to infer.

Second-order unification is still undecidable, but their are semi-algorithms that often work well in some cases.

This method was proposed by Pfenning [1988].

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Implicit	type argu	ments			

Derived from this solution, one can add decorations to let-bindings to indicate that some type arguments are left implicit.

Then, every occurrence of such a variable automatically adds type applications holes at the corresponding positions and type parameters will be inferred using second-order unification.

Other type applications must be passed explicitly.

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Bidirectio	nal type	inference			

What makes type-checking easy is that typing rules have an algorithmic reading. This implies that they are syntax directed, but also that judgments can be read as functions where some arguments are inputs and others are output.

Typically, Γ and a would be inputs and T is an ouput in the judgement $\Gamma \vdash a:T$.

However, althiough the rules for simply-typed λ -calculus are syntax directed they do not have an algorithmic reading;

The rule for abstraction is

```
\frac{\Gamma, x: T_0 \vdash a: T}{\Gamma \vdash \lambda x. a: T_0 \to T}
```

Then T_0 is used both as input and output in the premise.

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Bidirecti	onal type	inference			

However, in some cases, the type of the function may be known, *e.g.* when the function is an argument to an expression of a known type. In such cases, it suffices to check the proposed type is indeed correct.

Formally, the typing judgment $\Gamma \vdash a:T$ may be split into two judgments $\Gamma \vdash a \downarrow T$ to check that a may be assigned the type T and $\Gamma \vdash a \uparrow T$ to infer the type T of a.

Introduction S	imple types	Core ML	Type annotations	Recursive Types	System F
Bidirection	al type	inference		simple	types
$\frac{T = \Gamma(x)}{\Gamma \vdash x \Uparrow T}$		T ₀ ⊢ a	^{Арр-1} Г⊢а1 ∦Т2 - Г⊢	→ T ₁ 「⊢ a · a ₁ a ₂ ↑ T ₁	2 ↓ T 2

IntroductionSimple typesCore MLType annotationsRecursive TypesSystem FBidirectional type inferencesimple types
$$Var-I$$
 $T = \Gamma(x)$ $Abs-C$ $\Gamma, x : T_0 \vdash a \Downarrow T$ $\Gamma \vdash a_1 \Uparrow T_2 \rightarrow T_1$ $\Gamma \vdash a_2 \Downarrow T_2$ $\Gamma \vdash x \Uparrow T$ $\Gamma \vdash \lambda x. a \Downarrow T_0 \rightarrow T$ $\Gamma \vdash a_1 \Uparrow T_2 \rightarrow T_1$ $\Gamma \vdash a_2 \Downarrow T_2$ $I-C$ $\Gamma \vdash a \Uparrow T$ $\Gamma \vdash a \updownarrow T$ $\Gamma \vdash a \Downarrow T$ Γ

Checking mode can use inference mode.

Introduction	Simple types	Core ML	Type annotations	Recursive 1	Гурев	System F
Bidirecti	onal type	inference		6	imple	types
			App-I			
$\frac{V \text{ ar-l}}{\Gamma = \Gamma(x)}$		$T_0 \vdash a \Downarrow T$ $(a \Downarrow T_0 \to T)$	$\Gamma \vdash a_1 \Uparrow T_2$	2 → T ₁ ⊢ a ₁ a ₂ ↑		↓ <i>T</i> ₂
I-C		Annot-I	, i		1	
Г н	a ≬ T a ↓ T	$\frac{\Gamma \vdash a \Downarrow T}{\Gamma \vdash (a:T) \Uparrow}$	T			
		5e inference m				

Annotations turn inference mode into checking mode.

Introduction	Simple types	Core ML	Type annotations	Recursive Types	System F
Bidirect	ional type	inference		simple	e types
Var-I	Abs-C		App-I		
<u></u> <i>Τ</i> = Γ(> Γ ⊢ × ↑	<u> </u>	T ₀ ⊢ a ↓ T <.a ↓ T ₀ → T		$\rightarrow T_1 \qquad \Gamma \vdash a_1 = a_1 = a_2 \uparrow T_1$	a ₂
I-С Г I	- a (T	Annot-I Г⊢а ↓ 7	Аbs-I Г,	x : T ₀ ⊢ a ↑ T	
۲ı	- a ↓ T	Γ⊢ (a:⊺) ≬	$T \qquad \Gamma \vdash \lambda$	$1x:T_0.a \uparrow T_0 \rightarrow$	Т
Checking mode can use inference mode.					

Annotations turn inference mode into checking mode.

Annotations on type abstractions enable the inference mode.

Example: Let T be $(T_1 \rightarrow T_1) \rightarrow T_2$. and T be f:T

$$Var-I \xrightarrow{\Gamma \mapsto f \Uparrow T} \frac{\overline{\Gamma, x : T_1 \vdash x \Uparrow T_1}}{\Gamma, x : T_1 \vdash x \Downarrow T_1} e^{-I}$$

$$App-I \xrightarrow{\Gamma \vdash f \Uparrow T} \frac{\Gamma \vdash f (\lambda x. x) \Downarrow T_1 \to T_1}{\Gamma \vdash \lambda x. x \Downarrow T_1 \to T_1} Abs-C$$

$$Abs-C \xrightarrow{I-C} \frac{\Gamma \vdash f (\lambda x. x) \Uparrow T_2}{\varphi \vdash \lambda f : T. f (\lambda x. x) \Downarrow T \to T_2}$$

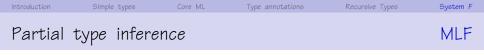
The method can be extended to deal with polymorphic types.

The idea is due to [Cardelli, 1993] and is still being improved [Dunfield, 2009]. However, it is quite complicated.

Predicative polymorphism is an interesting subcase where partial type inference can be reduced to typing constraints under a mixed prefix. Unfortunately, predicative polymorphism is too restrictive for programing languages (See [Rémy, 2005]).

A simpler approach proposed by Pierce and Turner [2000] and improved by Odersky et al. [2001] is perform bidirectional type inference but only from a small context surounding each node.

Interestingly, bidrectional type inference can easily be extended to work in the presence of subtyping.



MLF follows another approach that amounts to performing first-order unification of higher-order types.

- only parameters of functions that are used polymorphically need to be annotated.
- type abstractions and type annotation are always implicit

However, MLF goes beyond System F: for the purpose of type inference, it introduces richer types that enable to write "more principal types", but that are also harder to read. See [Rémy and Yakobowski, 2008].

The type inference method for MLF can be seen as a generalization of type constraints for ML to handle polymorphic types—still with first-order unification.

Bibliography I

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- Richard Bird and Lambert Meertens. Nested datatypes. In International Conference on Mathematics of Program Construction (MPC), volume 1422 of Lecture Notes in Computer Science, pages 52-67. Springer, 1998.
- Michael Brandt and Fritz Henglein. Coinductive axiomatization of recursive type equality and subtyping. Fundamenta Informaticæ, 33: 309-338, 1998.
 - Luca Cardelli. An implementation of $f_{j} ::$ Technical report, DEC Systems Research Center, 1993.
- Henry Cejtin, Matthew Fluet, Suresh Jagannathan, and Stephen Weeks. The MLton compiler, 2007.

Bibliography II

- Joshua Dunfield. Greedy bidirectional polymorphism. In ML '09: Proceedings of the 2009 ACM SIGPLAN workshop on ML, pages 15–26, New York, NY, USA, 2009. ACM. ISBN 978-1-60558-509-3. doi: http://doi.acm.org/10.1145/1596627.1596631.
- Fritz Henglein. Type inference with polymorphic recursion. ACM Transactions on Programming Languages and Systems, 15(2): 253-289, April 1993.
- J. Roger Hindley. The principal type-scheme of an object in combinatory logic. Transactions of the American Mathematical Society, 146:29-60, 1969.
 - Gérard Huet. Résolution d'équations dans des langages d'ordre 1, 2, ...,
 ω. PhD thesis, Université Paris 7, September 1976.
- Mark P. Jones. Typing Haskell in Haskell. In Haskell workshop, October 1999.

Bibliography III

- Assaf J. Kfoury, Jerzy Tiuryn, and Pawel Urzyczyn. ML typability is DEXPTIME-complete. In Colloquium on Trees in Algebra and Programming, volume 431 of Lecture Notes in Computer Science, pages 206–220. Springer, May 1990.
- Harry G. Mairson. Deciding ML typability is complete for deterministic exponential time. In ACM Symposium on Principles of Programming Languages (POPL), pages 382-401, 1990.
- David McAllester. A logical algorithm for ML type inference. In Rewriting Techniques and Applications (RTA), volume 2706 of Lecture Notes in Computer Science, pages 436–451. Springer, June 2003.
- Robin Milner. A theory of type polymorphism in programming. Journal of Computer and System Sciences, 17(3):348–375, December 1978.
- Alan Mycroft. Polymorphic type schemes and recursive definitions. In International Symposium on Programming, volume 167 of Lecture Notes in Computer Science, pages 217–228. Springer, April 1984.

Bibliography IV

- Martin Odersky, Martin Sulzmann, and Martin Wehr. Type inference with constrained types. Theory and Practice of Object Systems, 5(1): 35-55, 1999.
- Martin Odersky, Matthias Zenger, and Christoph Zenger. Colored local type inference. In ACM Symposium on Principles of Programming Languages (POPL), pages 41–53, 2001.
- Chris Okasaki. Purely Functional Data Structures. Cambridge University Press, 1999.
- Simon Peyton Jones and Mark Shields. Lexically-scoped type variables. Manuscript, April 2004.

Frank Pfenning. Partial polymorphic type inference and higher-order unification. In LFP '88: Proceedings of the 1988 ACM conference on LISP and functional programming, pages 153–163, New York, NY, USA, 1988. ACM. ISBN 0-89791-273-X. doi: http://doi.acm.org/10.1145/62678.62697.

Bibliography V

- Benjamin C. Pierce and David N. Turner. Local type inference. ACM Transactions on Programming Languages and Systems, 22(1):1-44, January 2000.
- ▷ François Pottier. Notes du cours de DEA "Typage et Programmation", December 2002.
- François Pottier and Didier Rémy. The essence of ML type inference. In Benjamin C. Pierce, editor, Advanced Topics in Types and Programming Languages, chapter 10, pages 389–489. MIT Press, 2005.
- François Pottier and Didier Rémy. The essence of ML type inference. Draft of an extended version. Unpublished, September 2003.
- Didier Rémy. Simple, partial type-inference for System F based on type-containment. In Proceedings of the tenth International Conference on Functional Programming, September 2005.

Bibliography VI

Didier Rémy and Boris Yakobowski. Efficient Type Inference for the MLF language: a graphical and constraints-based approach. In The 13th ACM SIGPLAN International Conference on Functional Programming (ICFP'08), pages 63–74, Victoria, BC, Canada, September 2008. doi: http://doi.acm.org/10.1145/1411203.1411216.

Christian Skalka and François Pottier. Syntactic type soundness for HM(X). In Workshop on Types in Programming (TIP), volume 75 of Electronic Notes in Theoretical Computer Science, July 2002.
 Robert Endre Tarjan. Efficiency of a good but not linear set union algorithm. Journal of the ACM, 22(2):215-225, April 1975.

Andrew Tolmach and Dino P. Oliva. From ML to Ada: Strongly-typed language interoperability via source translation. Journal of Functional Programming, 8(4):367-412, July 1998.

Bibliography VII

▷ J. B. Wells. Typability and type checking in system F are equivalent and undecidable. Annals of Pure and Applied Logic, 98(1-3): 111-156, 1999.