

Type Inference, Higher Order Algebra, and Lambda Calculus

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The Topics

Type Inference: how to find the possible type(s) of expressions, without explicit typing

Higher Order Algebra: a number of laws that the higher order functions like `map`, `fold` etc. obey

Lambda Calculus: a formal calculus for functions and how to compute with them

Type Inference

We have seen that the F# compiler can find types for expressions, and declared values:

```
let rec length l =  
  match l with  
  | []      -> 0  
  | _::xs  -> 1 + length xs
```

```
length : 'a list -> int
```

As we have mentioned, the *most general* type is always found

How can the compiler do this?

There is an interesting theory behind F#-style type inference

To infer means “to prove”, or “to deduce”

A type system is a *logic*, whose statements are of form “under some assumptions A , expression e has type τ ”

Often written “ $A \vdash e : \tau$ ”

To infer a type means to *prove* that a statement like above is true

A *type inference algorithm* finds a type if it exists: it is thus a *proof search algorithm*

Such an algorithm exists for F#'s type system

Logical Systems

A logical system is given by a set of *axioms*, and *inference rules* over a language of *statements*

A statement is true in the logic if it can be proved in a finite number of steps using these rules

Each inference rule has a number of *premises* and a *conclusion*

Often written on the form

$$\frac{\text{premise 1} \ \dots \ \text{premise } n}{\text{conclusion}}$$

Logical Systems

An example of an inference rule (modus ponens in propositional logic):

$$\frac{P \quad P \implies Q}{Q}$$

Hindley-Milner's Type System

F#'s type system extends a simpler type system known as *Hindley-Milner's type system* (HM)

This system was first invented around 1970

The typing statements have the form $A \vdash e : \tau$, where A is a set of typings for variables, e is an expression, and τ is a type

Example: $\{x : \alpha, f : \alpha \rightarrow \beta\} \vdash f x : \beta$

The type system of F# is basically the HM type system, with some extensions

Hindley-Milner Inference Rules

A selection of rules from the HM inference system:

$$A \cup \{x : \tau\} \vdash x : \tau \quad [VAR]$$

$$\frac{A \cup \{x : \sigma\} \vdash e : \tau}{A \vdash \lambda x. e : \sigma \rightarrow \tau} \quad [ABS]$$

$$\frac{A \vdash e : \sigma \rightarrow \tau \quad A \vdash e' : \sigma}{A \vdash e e' : \tau} \quad [APP]$$

$$\frac{A \vdash e : \forall \alpha. \tau}{A \vdash e : \tau[\sigma/\alpha]} \quad [SPEC]$$

(You don't need to learn this: I'm showing it only to let you know what an inference system might look like)

Inference Algorithm

There is a classical algorithm for type inference in the HM system

Called *algorithm \mathcal{W}*

Basically a systematic and efficient way to infer types

The algorithm uses *unification*, which is basically a symbolic method to solve equations

It has been proved that algorithm \mathcal{W} always yields a most general type for any typable expression

“Most general” means that any other possible type for the expression can be obtained from the most general type by instantiating its type variables

A Type Inference Example

Define

```
let rec length l =  
  match l with  
  | []      -> 0  
  | x::xs  -> 1 + length xs
```

Derive the most general type for `length`!

See next eight slides for how to do it ...

Type inference can be seen as equation solving: every declaration gives rise to a number of “type equations” constraining the types for the identifiers

These equations can be solved to find the types

For a declaration we basically do this:

- Find a typing for the left-hand side (LHS), using the typing rules
- Same for the RHS
- Ensure that LHS and RHS have the same type

If we succeed, then we have found a typing for the declared entity. If not, then there is a type error somewhere

Setting up the Equations (I)

In our example, we already know:

```
0, 1 : int
(+)  : 'n -> 'n -> 'n, 'n some numerical type
[]   : 'a list
(::) : 'b -> 'b list -> 'b list
```

Note the different type variable names, to make sure the types are independent

These typings will stay as they are throughout the inference process

Setting up the Equations (II)

We give the identifiers the following initial types:

```
length : 'c  
l       : 'd  
x       : 'e  
xs      : 'f
```

Each identifier is given a totally independent type. As the type inference proceeds, their types will become more and more constrained in order to fulfil the typing rules

When we're done, the typing of `length` can be read off the table

Typing the Left-hand Side

LHS:

`length l = ...`

`length` must have a function type, whose argument type is the type of `l`.

Thus

`'c = 'd -> 'g`

where `'g` is a new type variable. We also obtain

`length l : 'g`

So the type of the LHS is `'g`

Typing the Right-hand Side

The RHS is a `match` expression

They have the following typing rules:

- The matched expression, and all the patterns, must have the same type
- The results must all have the same type
- The type of the `match` expression is the type of the results

We check these rules next

Typing the Matched Expression, and the Patterns

First we check that the pattern $x :: xs$ is well-typed. This requires:

```
'e = 'b, 'f = 'b list
```

With these typings we obtain

```
x :: xs : 'b list
```

Now l , $[]$, $x :: xs$ should have the same type. This implies

```
'd = 'a list = 'b list
```

(which requires that $'a = 'b$). Since $\text{length} : 'd \rightarrow 'g$, we now have

```
length : 'b list  $\rightarrow$  'g
```

Typing the Results, and the Right-hand Side

0, and `1 + length xs` should have the same type, which then becomes the type of the RHS

We have `0 : int`

What about `1 + length xs`? We have `xs : 'b list`, so `length xs` is well-typed with type `'g`. Thus, `1 + length xs` is well-typed if:

`'g = 'n, 'n = int`

This implies

`length : 'b list -> int`

We also obtain that LHS and RHS both have type `int`. We're done!

Most General Type

The type inferred for `length` is its *most general type*

This is since we were careful not to make any stronger assumptions than necessary about any types in each step of the inference

Another Type Inference Exercise

Find the most general type for `int_halfve`, defined by:

```
let rec int_halfve a l u =  
  if u = l+1 || a.[l] = 0.0 || a.[u] = 0.0 then (l,u)  
  else let h = (l+u)/2 in  
    if a.[h] > 0 then int_halfve a l h  
    else int_halfve a h u
```

Higher Order Algebra

Higher order functions like `map`, `fold`, `>>`, ... obey certain *laws*

These laws can be compared to laws for arithmetical operators, like

$$x + (y + z) = (x + y) + z$$

They can be used to transform programs, e.g., optimizing them

They also help understanding the functions better

Some Laws involving List.map

`map id = id`, where `id = fun x -> x` (the identity function)

`map (g >> f) = map g >> map f`

`map f >> tail = tail >> map f`

`map f >> reverse = reverse >> map f`

`map f (xs @ ys) = map f xs @ map f ys`

(Writing `map`, etc. for `List.map`, etc.)

Some Laws involving List.filter

`filter p >> reverse = reverse >> filter p`

`filter p (xs @ ys) = filter p xs @ filter p ys`

`map f >> filter p = filter (f >> p) >> map f`

A Property of Fold

If op is associative and if e is left and right unit element for op , then, for all lists xs :

`foldBack op xs e = fold op e xs`

What Can Laws Like This Be Used For?

A simple example: rewriting to optimize code

```
reverse >> filter p >> map f >> reverse =  
filter p >> reverse >> map f >> reverse =  
filter p >> map f >> reverse >> reverse =  
filter p >> map f >> id =  
filter p >> map f
```

since obviously

```
reverse >> reverse = id
```

How to Prove the Laws

Mathematical laws need mathematical proofs

How can the laws for higher-order functions be proved?

We'll exemplify with the law

$$\text{map } f \text{ (xs @ ys)} = \text{map } f \text{ xs @ map } f \text{ ys}$$

- First, informal reasoning (to motivate why the law holds)
- Then, a formal proof using induction over lists

An Informal Proof

Let $xs = [x_1, \dots, x_m]$, $ys = [y_1, \dots, y_n]$

Then

$$\begin{aligned} \text{map } f \ ([x_1, \dots, x_m] \ @ \ [y_1, \dots, y_n]) &= \text{map } f \ ([x_1, \dots, x_m, y_1, \dots, y_n]) \\ &= [f \ x_1, \dots, f \ x_m, f \ y_1, \dots, f \ y_n] \\ &= [f \ x_1, \dots, f \ x_m] \ @ \ [f \ y_1, \dots, f \ y_n] \\ &= \text{map } f \ [x_1, \dots, x_m] \ @ \\ &\quad \text{map } f \ [y_1, \dots, y_n] \end{aligned}$$

That is,

$$\text{map } f \ (xs \ @ \ ys) \ = \ \text{map } f \ xs \ @ \ \text{map } f \ ys$$

Q.E.D.

An Formal Proof

If you really want to be sure . . .

A proof by **induction**

The proof will be over the **structure of lists**

It will use the **recursive definitions** of @ and map

Proof by Induction

Have you ever performed proofs by induction? (You should have...)

They prove properties that hold for *all non-negative integers*

For instance, $\forall n. \sum_{i=0}^n i = n(n+1)/2$

Exercise: prove this property by induction!

But first, let's check out next slide ...

The Induction Principle for Natural Numbers

Goal: show that the property P is true for all natural numbers (whole numbers ≥ 0)

Proof by induction goes like this:

1. Show that P holds for 0 (the *base case*)
2. Show, for all natural numbers n , that if P holds for n then P holds also for $n + 1$ (the *induction step*)
3. Conclude that P holds for all n

To prove 2 one typically assumes that $P(n)$ is true (the *induction hypothesis*), then shows that $P(n + 1)$ follows

Why does Induction over the Natural Numbers Work?

The set of natural numbers \mathbf{N} is an *inductively defined set*

\mathbf{N} is defined as follows:

- $0 \in \mathbf{N}$
- $\forall x. x \in \mathbf{N} \implies s(x) \in \mathbf{N}$ (the *successor* of x , i.e., $x + 1$)

$$\begin{array}{ccccccc} 0 & \rightarrow & s(0) & \rightarrow & s(s(0)) & \rightarrow & s(s(s(0))) & \rightarrow & \dots \\ 0 & & 1 & & 2 & & 3 & & \dots \end{array}$$

Proofs by induction follow the structure of the inductively defined set!

The Inductively Defined Set of Lists

Inductively defined sets are typically sets of *infinitely* many *finite* objects

The set `'a list` of (finite) lists with elements of type `'a`:

1. `[] ∈ 'a list`
2. `x ∈ 'a ∧ xs ∈ 'a list ⇒ x :: xs ∈ 'a list`

Note similarity with the set of natural numbers!

Also cf. the following type declaration (in “pseudo”-F#):

```
type 'a list = [] | (:::) of 'a * 'a list
```

An Induction Principle for Lists

Proof by induction for (finite) lists goes like this:

1. Show that P holds for `[]`
2. Show, for all finite lists `xs ∈ 'a list` and all possible list elements `x ∈ 'a`, that if P holds for `xs` then P holds also for `x :: xs`
3. Conclude that P holds for all finite lists in `'a list`

The Formal Proof

Now let's formally prove our equality

Prove that, for all xs, ys, f holds that:

$$\text{map } f \text{ (} xs @ ys \text{)} = \text{map } f \text{ } xs @ \text{map } f \text{ } ys$$

What induction hypothesis to use? This is often the tricky question!

General rule: look at the function definitions, and try to formulate the induction hypothesis so it matches the recursive structure!

Function Definitions

From the definitions of $(@)$ and `map` we obtain:

$$\begin{aligned} [] @ ys &= ys \\ (x :: xs) @ ys &= x :: (xs @ ys) \end{aligned}$$
$$\begin{aligned} \text{map } f [] &= [] \\ \text{map } f (x :: xs) &= f x :: \text{map } f xs \end{aligned}$$

(“Mathematical” case-by-case versions of the function definitions)

$@$ recurses over its first argument (xs in the statement to prove)

Thus, let's do the induction over xs

Induction Hypothesis

This is then our induction hypothesis:

$$P(xs) = \text{map } f \ (xs \ @ \ ys) = \text{map } f \ xs \ @ \ \text{map } f \ ys$$

If we can prove $\forall xs. P(xs)$, then we have proved that the law holds!

We will now prove the following:

1. $P([])$ (base case)
2. $\forall x. \forall xs. [P(xs) \implies P(x :: xs)]$ (induction step)

By the induction principle for lists, this will prove $\forall xs. P(xs)$

Base Case

$$P([]) = \text{map } f ([] @ ys) = \text{map } f [] @ \text{map } f ys$$

Assume any ys, f

Let's show that the LHS equals the RHS:

$$\begin{aligned} \text{LHS} &= \text{map } f ([] @ ys) \\ &= \text{map } f ys \\ \text{RHS} &= \text{map } f [] @ \text{map } f ys \\ &= [] @ \text{map } f ys \\ &= \text{map } f ys \end{aligned}$$

Thus $\text{LHS} = \text{RHS}$, and $P([])$ holds

Induction step

We want to prove

$$P(x :: xs) = \text{map } f ((x :: xs) @ ys) = \text{map } f (x :: xs) @ \text{map } f \text{ } ys$$

We are allowed to use $P(xs)$ in the proof. Assume any ys, f . Then,

$$\begin{aligned} \text{LHS} &= \text{map } f ((x :: xs) @ ys) \\ &= \text{map } f (x :: (xs @ ys)) \\ &= f \ x :: \text{map } f (xs @ ys) \\ &= (\text{induction hypothesis}) \\ &= f \ x :: (\text{map } f \ xs @ \text{map } f \ ys) \\ &= (f \ x :: \text{map } f \ xs) @ \text{map } f \ ys \\ &= \text{map } f (x :: xs) @ \text{map } f \ ys \\ &= \text{RHS} \end{aligned}$$

Conclusion

We showed the base case $P([])$, and the induction step $P(xs) \implies P(x :: xs)$

We can thus conclude that $\forall xs. P(xs)$

That is, the law holds

Bird-Meertens Formalism

The identities shown belong to an *algebra of list functions*

This is known as the *Bird-Meertens Formalism*

The idea of Bird and Meertens was to do program development by:

- making a *specification* of the program, using the list primitives, and
- using the identities to *transform* the specification into an efficient implementation

This attempt has not been overly successful in general, but I think there are niches where the method can be applied

In particular, it has been proposed for programming of parallel computers

Lambda Calculus

Formal calculus

Invented by logicians around 1930 (Curry, Schönfinkel, and others)

Formal syntax for functions, and function application

Gives a certain “computational” meaning to function application

Theorems about reduction order (which possible subcomputation to execute first)

This is related to call-by-value/call-by-need

Several variations of the calculus



H. B. Curry



M. Schönfinkel

The Simple Untyped Lambda Calculus

The calculus consists of a *language*, and *equivalences* on expressions in the language. A term in the language is:

- a *variable* x ,
- a *lambda-abstraction* $\lambda x.e$, or
- an *application* $e_1 e_2$

Some examples:

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

Any term can be applied to any term, no concept of (function) types

Syntax

Function application binds strongest: $\lambda x.e_1 e_2 = \lambda x.(e_1 e_2) \neq (\lambda x.e_1) e_2$

Function application is left associative: $e_1 e_2 e_3 = (e_1 e_2) e_3$

Lambda Calculus Syntax and Functional Programming

Syntax elements from the lambda calculus have been adopted by higher order functional languages, in particular:

- Function expressions `(fun x -> e)`, from $\lambda x.e$
- Function application syntax, and currying: `f e1 e2`

Untyped Lambda Calculus with Constants

We can extend the syntax with constants, for instance:

1, 17, +, [], ::

We can then form terms closer to usual functional languages, like

$17 + x$ $\lambda x.(x + y)$ $\lambda l.\lambda x.(l :: x)$

Functional language compilers often first translate into an intermediate form, which essentially is a lambda calculus with constants

Equivalences

Some lambda-expressions are considered equivalent ($e_1 \equiv e_2$)

Rule 1: change of name of bound variable gives an equivalent expression (*alpha-conversion*)

So $\lambda x.(x\ x) \equiv \lambda y.(y\ y)$

Quite natural, right? If we change the name of the formal parameter, the function should still be the same

Example: in F#, `fun x -> x` and `fun y -> y` define the same function

Variable Capture

However, beware of *variable capture*:

$$\lambda x. \lambda y. x \neq \lambda y. \lambda y. y$$

Renaming must avoid name clashes with locally bound variables

Precisely the same problem appears in programming languages:

```
let f x = let g y = x + y in ...
```

Here we cannot change x into y without precautions. However, OK if we rename y in g to z first:

```
let f x = let g z = x + z in ... =>
```

```
let f y = let g z = y + z in ...
```

Same trick is used in lambda calculus: $\lambda x. \lambda y. x \equiv \lambda x. \lambda z. x \equiv \lambda y. \lambda z. y$

Beta-reduction

A lambda abstraction applied to an expression can be *beta-reduced*:

$$(\lambda x. x + x) 9 \rightarrow_{\beta} 9 + 9$$

Beta-reduction means substitute actual argument for symbolic parameter in function body

A formal model for what happens when *a function is applied to an argument*

Works also with symbolic arguments:

$$(\lambda x. x + x) (\lambda x. y z) \rightarrow_{\beta} (\lambda x. y z) + (\lambda x. y z)$$

Like *inlining* done by optimizing compilers

Variable Capture

However, again beware of variable capture:

$$(\lambda x. \lambda y. (x + y)) y \not\rightarrow_{\beta} \lambda y. (y + y)$$

The fix is to first rename the bound variable y :

$$(\lambda x. \lambda y. (x + y)) y \equiv (\lambda x. \lambda z. (x + z)) y \rightarrow_{\beta} \lambda z. (y + z)$$

The same thing can happen when inlining functions. Example:

```
let f x = let g y = x + y in ...  
let h y = f (y + 3)
```

If we want to inline the call to `f` in `h`, then `g`'s argument must first be renamed:

```
...let g z = x + z in ...
```

```
let h y = f (y + 3) =>  
let h y = let g z = (y + 3) + z in ...
```

Some Encodings

Many mathematical concepts can be *encoded* in the (untyped) lambda-calculus

That is, they can be translated into the calculus

For instance, we can encode the *boolean constants*, and a *conditional* (functional if-then-else):

$$TRUE = \lambda x.\lambda y.x$$

$$FALSE = \lambda x.\lambda y.y$$

$$COND = \lambda p.\lambda q.\lambda r.(p\ q\ r)$$

Exercise: make these encodings in F#!

An example of how *COND* works:

$$\begin{aligned} \text{COND TRUE } A B &\rightarrow_{\beta} (\lambda p.\lambda q.\lambda r.(p q r)) (\lambda x.\lambda y.x) A B \\ &\rightarrow_{\beta} (\lambda q.\lambda r.((\lambda x.\lambda y.x) q r)) A B \\ &\rightarrow_{\beta} (\lambda r.((\lambda x.\lambda y.x) A r)) B \\ &\rightarrow_{\beta} (\lambda x.\lambda y.x) A B \\ &\rightarrow_{\beta} \lambda y.A B \\ &\rightarrow_{\beta} A \end{aligned}$$

Try evaluating *COND FALSE A B* yourself!

Boolean connectives (and, or) can also be encoded

As well as lists, integers, . . . Even *recursion* can be encoded as a lambda expression

Actually *anything you can do in a functional language!*

This means that *any functional program* can be translated into the lambda calculus

Thus, lambda calculus serves as a general model for functional languages

Nontermination

Consider this expression:

$$(\lambda x. x x) (\lambda x. x x)$$

What if we beta-reduce it?

$$(\lambda x. x x) (\lambda x. x x) \rightarrow_{\beta} (\lambda x. x x) (\lambda x. x x)$$

Whoa, we got back the same! Scary ...

Clearly, we can reduce ad infinitum

The lambda-calculus thus contains *nonterminating* reductions

Reduction Strategies

Any application of a lambda-abstraction in an expression can be beta-reduced

Each such position is called a *redex*

An expression can contain several redexes

Can you find all redexes in this expression?

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

Try reduce them in different orders!

Does the order of reducing redexes matter?

Well, yes and no:

Theorem: *if two different reduction orders of the same expression end in expressions that cannot be further reduced, then these expressions must be the same*

However, we can have potentially infinite reductions:

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

Reducing the “outermost” redex yields y

But the innermost redex can be reduced infinitely many times – nontermination!

So the order *does* matter, as regards termination anyway!

Normal Order Reduction

There is something called “normal order reduction” in the lambda calculus

It is a strategy to select which redex to reduce next

Normal order reduction corresponds to lazy evaluation, or call by need

Theorem: *if there is a reduction order that terminates, then normal order reduction terminates*

For functional languages, this means that lazy evaluation always is the “best” in the sense that it terminates whenever the program terminates with some other reduction strategy, like call by value