

Dependent Types in Programming

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TYPES summer school
Giens, 5 September 2002

Constructive mathematics and computer programming – the original paradigm

Curry-Howard for programming:

a	::	A
element	belongs to	set/type
proof	proves	proposition
program	satisfies	specification

Martin-Löf type theory:

- a functional language with dependent types where all programs terminate;

- a specification language including predicate logic;
- a full-scale constructive set theory – a “ZF” for constructive mathematics!

Example: sorting

$$\text{SortProp} = \forall xs :: [\text{Nat}]. \exists ys :: [\text{Nat}]. \\ \text{Sorted } ys \wedge \text{Perm } xs \text{ } ys$$

Prove this proposition!

```
sortProof :: SortProp
```

Extract a program

```
sortProg :: [Nat] -> [Nat]
```

with its proof

```
sortProgProof  ::  ∀xs :: [Nat].  
                Sorted (sortProg xs) ∧  
                Perm xs (sortProg xs)
```

Program extraction

Set vs Prop. Distinguish between computationally relevant and irrelevant parts by using Set/Prop-distinction. The sorting proposition becomes eg

$$\prod xs :: [\text{Nat}]. \{ys :: [\text{Nat}] \mid \text{Sorted } ys \wedge \text{Perm } xs \text{ } ys\}$$

One-element types are not computationally relevant. Eg use that Sorted (and Perm) are decidable:

```
sorted :: [Nat] -> Bool
Sorted :: [Nat] -> Set
Sorted xs = T (sorted xs)
```

where

```
T :: Bool -> Set
```

```
T True  = Unit = ()
```

```
T False = Empty
```

Constructive mathematics and computer programming - what happened?

- 1979** - The paper “Constructive mathematics and computer programming”. Formation of Cornell and Chalmers type theory groups. Early implementations (NuPRL, GTT system).
- 1984** - The calculus of constructions. Logical frameworks. INRIA and Edinburgh groups. Implementations of intensional type theory.
- 1989** - The Logical Framework/TYPES consortium. Lego, Coq, Alf. Inductive definitions, pattern matching, records, ...

2002 Impressive progress. But dependent type theory has not (yet?) revolutionized programming.

The next 700 MLs

Extensions of the Hindley-Milner type system (polymorphic typed lambda calculus with recursive type and function definitions):

- Equality types in ML.
- ML's module system. Haskell's class system.
- Arrays. Sized types. Embedded ML.
- Metaprogramming. Meta-ML.
- Generic programming. PolyP, Generic Haskell.

- Specification language for testing. QuickCheck.

Dependent types in practical programming

Dependent types from the point of view of the functional programmer (ML, Haskell):

Cayenne Augustsson 1998.

DML Xi 1998, Xi and Pfenning 1999.

Series of workshops on DTP: Göteborg 1999, Ponte de Lima 2000, Schloss Dagstuhl 2001, ...

What is Cayenne?

Augustsson 1998: “Although dependent types have been used before in proof systems, e.g., [CH88], to our knowledge this is the first time that the full power of dependent types has been integrated into a *programming language*.”

- A lazy functional language with dependent types, similar to Agda, but intended to be used as a “real” programming language, like Haskell. Unlike Agda it has predefined types `Int`, `String`, etc.
- Intended to be used as a partial type theory with unrestricted recursion in type and function definitions.
- Type-checking undecidable, but nevertheless practical.

- Compiled by removing types and translating to LML, which has a compiler producing efficient code.

What is DML?

Xi and Pfenning 1999: “To our knowledge, no previous type system for a general purpose programming language such as ML has combined dependent types with features including datatype declarations, higher-order functions, general recursions, let-polymorphism, mutable references, and exceptions.”

- A strict functional language with dependent types. DML = Dependent ML.
- A conservative extension of ML. Translate to ML by removing dependent types.
- Decidable type checking is obtained by restricting types to depend on *index expressions*, eg arithmetic expressions, with decidable constraint solving.

Plan

1. Haskell classes and dependent records.
2. Dependently typed datastructures.
3. Testing and dependent types.
4. Generic programming and universal algebra.
5. Metaprogramming. Well-typed interpreters and partial evaluators.
6. Coping with general recursion.

Notation - logical framework

Inspired by Haskell, Cayenne, Agda.

here

$x :: a$

$a \rightarrow b$

$(x :: a) \rightarrow b$

$f\ c$

$\backslash x \rightarrow e$

(a, b)

$(x :: a, b)$

$(x :: a, y :: b)$

$r.x, r.y$

$(c, d), (x=c, y=d)$

$()$

Bool

Set

$T :: \text{Bool} \rightarrow \text{Set}$

other

$x : a$

$(a)b$

$(x:a)b, \prod x:a.b$

$f(c)$

$\lambda x.e, (x)e$

$a \times b$

$\sum x:a.b$

$\text{sig}\{x:a, y:b\}$

fst, snd

$\text{struct}\{x=c, y=d\}$

$N_1, 1, \text{Unit}$

$N_2, 2$

$\#, *, \text{Prop}$

Lift

Notation - datatypes

here

```
Nat  :: Set
Zero :: Nat
Succ :: Nat -> Nat
```

```
[a]
Vect a n
[]
x : xs
xs ++ ys
```

```
BT a
BST a
IsBST t
```

```
(==)
```

other

```
data Nat = Zero
         | Succ Nat
```

```
List a
an
Nil
x.xs, Cons x xs
append xs ys
```

```
BinaryTree a
BinarySearchTree a
T (isBST t)
```

```
eq
```

Also argument hiding, overloading, Haskell (-) notation for infix operations, etc.

1. Haskell classes and dependent records

The Eq-class in Haskell

```
class Eq a where  
    (==) :: a -> a -> Bool
```

```
instance Eq Bool where  
    (==) = eqBool
```

where

```
eqBool :: Bool -> Bool -> Bool
```

is defined by

```
eqBool True  True  = True
```

```
eqBool False False = True
eqBool _      _      = False
```


Eq with records

Haskell's class declaration corresponds to

```
Eq :: Set -> Set
```

```
Eq a = a -> a -> Bool
```

If we want to specify the name (==) of the operation we can instead use a record type:

```
Eq a = ((==) :: a -> a -> Bool)
```

(This is the Eq “class” in Cayenne.)

The instance declaration corresponds to

```
eqBool :: Eq Bool
```

or, as a record with one field,

```
((==) = eqBool) :: Eq Bool
```

Overloading via classes

Wadler's original purpose of Haskell-classes was to have a systematic approach to overloading. Having defined an instance `Eq a` we can simply use

```
(==) :: a -> a -> Bool
```

for the equality on `a`.

This is not captured by our dependent records. If

```
r :: Eq a
```

is a record, we have to write

```
r.(==) :: a -> a -> Bool
```

I will however in some future examples assume that we can hide the τ also when we work in dependent type theory.

Eq with properties

The Eq-record with properties (decidable setoids = “datoids”):

```
Eq a = ((==) :: a -> a -> Bool,  
        ref  :: (x :: a)  
              -> T (x == x),  
        sym  :: (x,y :: a)  
              -> T (x == y)  
              -> T (y == x),  
        tra  :: (x,y,z :: a)  
              -> T (x == y) -> T (y == z)  
              -> T (x == z)  
    )
```

Deriving equality

What about writing a function

`eq :: (a :: Set) -> Eq a ?`

In total type theory this means that we should define a decidable equality for all `a :: Set`. But equality is not decidable for all sets! However, we could have

`eq :: (a :: EqSet) -> Eq a`

where `EqSet` is a universe of sets for which we can “derive” decidable equality (cf ML’s equality types). This leads us towards “generic programming” – more later.

Subclasses

Records are first-class citizens in dependent type theory. This helps us model some further class-related language constructs.

In Haskell `Ord` is a subclass of `Eq`. A simplified version:

```
class Eq a => Ord a where
  (<) :: a -> a -> Bool
```

In dependent type theory this corresponds to

```
Ord a = (r    :: Eq a,
         (<) :: r.a -> r.a -> Bool
        )
```

List equality

In Haskell:

```
instance Eq a => Eq [a] where
  []      == []      = True
  []      == (y:ys) = False
  (x:xs) == []      = False
  (x:xs) == (y:ys) = x == y && xs == ys
```

In dependent type theory we write

```
listEq :: (a :: Set) -> Eq a -> Eq [a]
```

ie essentially a function


```
listEq :: (a :: Set) -> (a -> a -> Bool)
        -> [a] -> [a] -> Bool
```

```
listEq f [] [] = True
```

```
listEq f [] (y:ys) = False
```

```
listEq f (x:xs) [] = False
```

```
listEq f (x:xs) (y:ys) = f x y && listEq xs ys
```

2. Dependently typed datastructures

The zip-function

Haskell library function

```
zip :: [a] -> [b] -> [(a,b)]
```

```
zip [] [] = []
```

```
zip (x:xs) (y:ys) = (x,y) : zip xs ys
```

```
zip _ _ = []
```

exceptional cases when lists are of unequal length.

Vectors

```
Vect :: Set -> Nat -> Set
```

`Vect a n` is the set of lists of length `n`, ie the set of `n`-tuples. Then `zip` gets the type

```
zip :: (a,b :: Set) -> (n :: Nat)
      -> Vect a n -> Vect b n
      -> Vect (a,b) n
```

Note that vectors can be defined either inductively (as an “inductive family”)

```
Nil  :: (a :: Set) -> Vect a Zero
Cons :: (a :: Set) -> (n :: Nat)
```

```
-> a -> Vect a n
-> Vect a (Succ n)
```

or recursively (using “large elimination”)

```
Vect a Zero      = ()
Vect a (Succ n) = (a, Vect a n)
```

Balanced binary trees

`Bal a h` is the set of balanced binary trees of height `h` and with `a`-elements in the nodes. “Balanced” here means “as in AVL-trees”.

```
Bal :: Set -> Nat -> Set
```

```
Empty  :: (a :: Set)  
        -> Bal a 0
```

```
BranchE :: (a :: Set) -> (h :: Nat)  
         -> a -> Bal a h -> Bal a h  
         -> Bal a (h+1)
```

```
BranchL :: (a :: Set) -> (h :: Nat)  
         -> a -> Bal a (h+1) -> Bal a h
```

```
      -> Bal a (h+2)
BranchR :: (a :: Set) -> (h :: Nat)
      -> a -> Bal a h -> Bal a (h+1)
      -> Bal a (h+2)
```

Binary search trees

```
Empty  :: (min,max :: Nat) -> T (min < max)
      -> BST min max
```

```
Branch :: (min,max,root :: Nat)
      -> T (min < root) -> T (root < max)
      -> BST min root -> BST root max
      -> BST min max
```

```
Branch min max root p q left right
```

is a binary search tree with bounds min and max, root root, left and right subtrees left and right, and p and q proofs that root is between min and max.

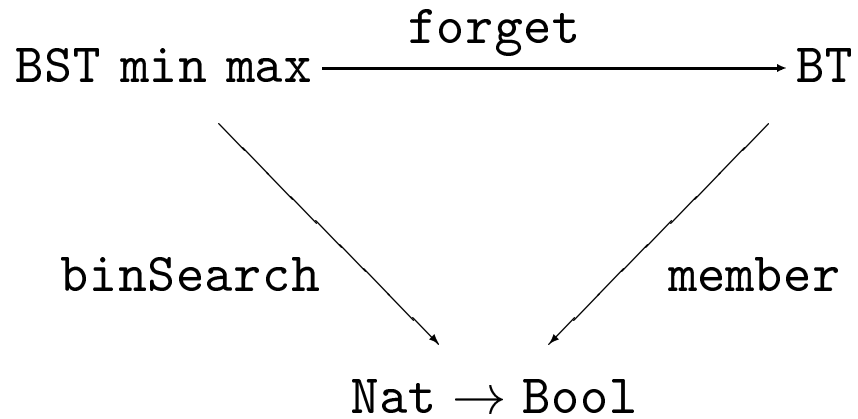
```
binSearch :: (min,max,key :: Nat)
```



```
-> T (min < key) -> T (key < max)
-> BST min max -> Bool
```

```
insert    :: (min,max,key :: Nat)
          -> T (min < key) -> T (key < max)
          -> BST min max
          -> BST min max
```

Correctness of binary search



where BT is the set of binary trees with natural numbers in the nodes:

Empty :: BT

Branch :: Nat -> BT -> BT -> BT

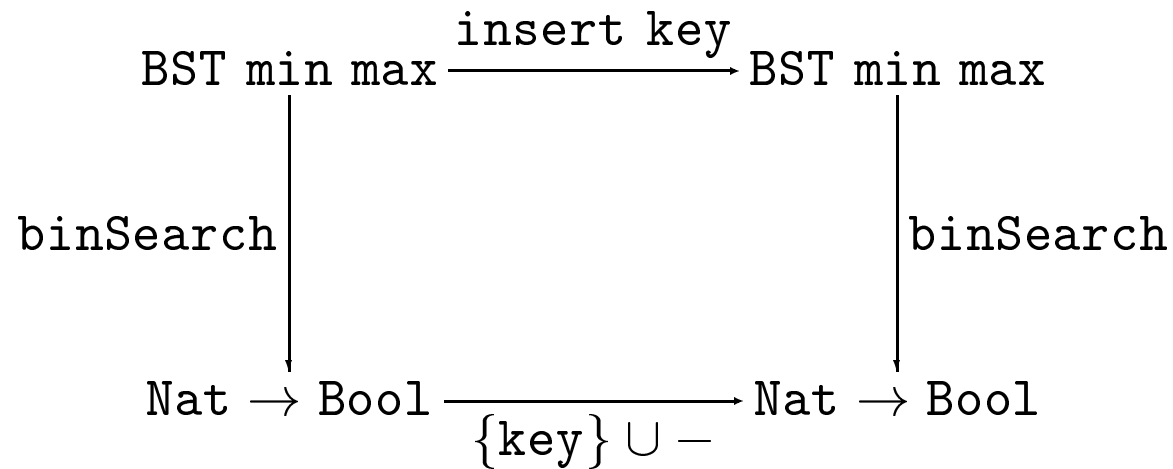
and

```
member :: BT -> Nat -> Bool
```

```
forget :: (min,max :: Nat) -> BST min max -> BT
```

are the obvious membership and binary search tree structure forgetting functions.

Correctness of insertion in binary search trees



$(U) :: (\text{Nat} \rightarrow \text{Bool}) \rightarrow (\text{Nat} \rightarrow \text{Bool})$

$\{-\} :: \text{Nat} \rightarrow \text{Nat} \rightarrow \text{Bool}$

The binary search tree property

```
isBST :: Nat -> Nat -> BT -> Bool
```

```
isBST min max Empty = min < max
```

```
isBST min max (Branch root left right)
```

```
= min < root && root < max
```

```
  && isBST min root left
```

```
  && isBST root max right
```

The dependently typed “integrated” representation is isomorphic to the “external” one:

$$\text{BST min max} \cong (t :: \text{BT}, T (\text{isBST min max } t))$$

3. Testing and dependent types

Random testing

QuickCheck (Claessen and Hughes 2000) is a tool for testing Haskell programs automatically. The programmer provides a specification of the program, in the form of properties which functions should satisfy, and QuickCheck then tests that the properties hold in a large number of randomly generated cases. Specifications are expressed in Haskell, using combinators defined in the QuickCheck library. QuickCheck provides combinators to define properties, observe the distribution of test data, and define test data generators.

From the ICFP programming contest

Tom Moertel wrote the following ...

Lesson 4. Test early, test often.

This one I got right. Early on I decided to invest a substantial portion of my time on correctness. I benefited from Haskell's wicked-powerful type system, which catches a lot of problems all by itself, and then I used QuickCheck, an automatic testing tool, to further automate away the pain of testing. Here are a few examples from my log that show how a tool like QuickCheck can be your best friend in a tight coding corner:

Thu 17:13 EDT. QuickCheck is revealing that something is going wrong with either the Parser or my Show instances. . . .

Thu 17:37 EDT. QuickCheck to the rescue! QuickCheck found a test case that falsified my RoundTrip property for Show->Parse->Show, and I was able to hand-feed that case to my parser to determine the error. . . .

Fri 12:19 EDT. My naive optimizer is done. Not so fast, QuickCheck spotted a corner case that causes the optimizer to discard untagged text at the end of a document. Oops.

QuickCheck found these problems and more, many that I wouldn't have found without a massive investment in test cases, and it did so quickly and easily. From now on, I'm a QuickCheck man!

QuickCheckable properties

QuickCheck can test conditional properties written

$$p\ x \implies q\ x$$

where

$$p, q :: D \rightarrow \text{Bool}$$

are decidable predicates written in Haskell.

In dependent type theory:

$$(x :: D) \rightarrow T\ (p\ x) \rightarrow T\ (q\ x)$$

The user writes a generator of random D-elements. QuickCheck uses this to check the conditional property for 100 cases, where only cases which pass p are counted. If a counterexample is found, QuickCheck stops and reports it.

Testing binary search

```
binSearch :: BT -> Nat -> Bool
```

is a Haskell version of binary search (now forgetting about bounds).

The correctness criterion written in QuickCheck's specification language is

```
isBST t ==> binSearch t key == member t key
```

and in dependent type theory:

```
(t :: BT) -> (key :: Nat) -> T (isBST t)  
-> T (binSearch t key == member t key)
```

Write a random generator for binary search trees!

Combining QuickCheck and dependent type theory

Methodology:

- Debug the goal by running QuickCheck. (The specification may as often be wrong as the program!)
- If successful refine the goal, until you get some subgoal that seems hard to prove. Run QuickCheck on this. And so on.

Hayashi used testing when doing proofs in his PX system already in the 1980-ies.

Other benefit of combining QuickCheck and dependent type theory:

- Prove surjectivity (“coverage”) properties of generators of random elements.

4. Generic programming and universal algebra

Generic equality

Recall that we said that we would like a function

```
eq :: (a :: EqSet) -> a -> a -> Bool
```

inside our language, which derives an equality for each set in EqSet - a universe of “equality sets” (like ML’s equality types).

We could even derive the equality with properties:

```
eq :: (a :: EqSet)
    -> ((==) :: a -> a -> Bool,
        ref  :: (x :: a)
                -> T (x == x),
```

```
sym  :: (x,y :: a)
      -> T (x == y)
      -> T (y == x),
tra  :: (x,y,z :: a)
      -> T (x == y) -> T (y == z)
      -> T (x == z)
)
```

Generic map

Another motivating example of generic programming.

$\text{map} :: (a \rightarrow b) \rightarrow [a] \rightarrow [b]$

is one of the most basic functions for list programming. But we have an analogous function for binary trees:

$\text{mapBT} :: (a \rightarrow b) \rightarrow \text{BT } a \rightarrow \text{BT } b$

In general we can define a generic map

$\text{map} :: (a \rightarrow b) \rightarrow D \ a \rightarrow D \ b$

where

$D :: \text{Set} \rightarrow \text{Set}$

is a unary datatype constructor. However, we cannot define `map` for an arbitrary such `D`, but only ones which are drawn from a suitable universe of “regular” datatypes.

Regular datatypes in PolyP

PolyP (Jansson and Jeuring, 1996 -) as in “polytypic” (= generic) programming, is an extension of Haskell.

Polytypic functions are defined by induction on a universe of codes for “regular datastructures”. These are unary type constructors of the form

$$DX = \mu Y. FXY$$

where F is a “pattern functor” built up from variables and constants by sum, product, and composition. Eg the type of lists of X 's is a regular datastructure defined by

$$[X] = \mu Y. () + (X, Y)$$

The functionality of PolyP can be simulated in dependent type theory.

Generic programming and universal algebra

We consider term algebras T_Σ for one-sorted signatures Σ .

A signature is just a list of arities:

`Sig = [Nat]`

and

`T :: Sig -> Set`

takes a signature and returns its term algebra.

Some signatures and their term algebras:

[]	Empty
[0,0]	Bool
[0,1]	Nat
[0,1,1]	[Bool]
[0,2]	BT ()

A generic size function

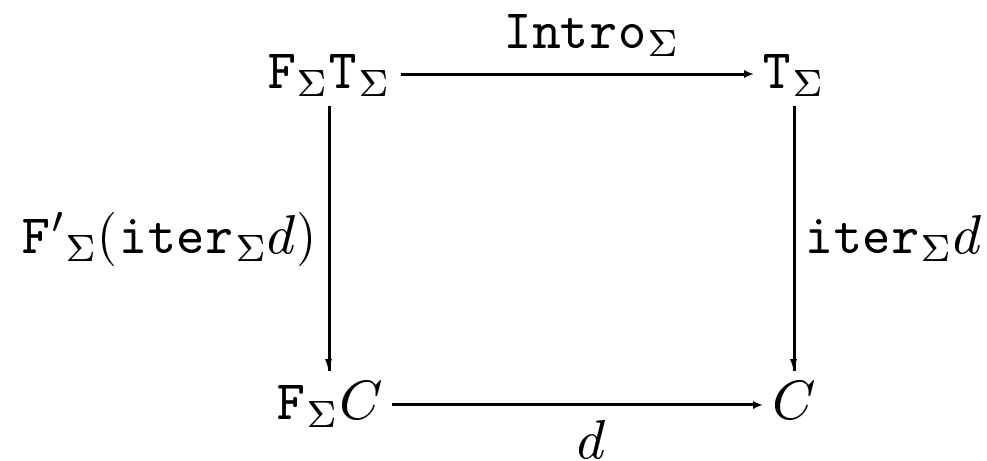
How to program

```
size :: (Sigma :: Sig) -> T Sigma -> Nat
```

by induction on Sig?

Generic formation, introduction, elimination, and equality rules

Use the well-know initial algebra diagram:



$T \quad :: \text{Sig} \rightarrow \text{Set}$

```
Intro :: (Sigma :: Sig)
      -> F Sigma (T Sigma) -> T Sigma
iter  :: (Sigma :: Sig) -> (C :: Set)
      -> (F Sigma C -> C) -> T Sigma -> C

iter Sigma d (Intro Sigma x)
= d (F' Sigma (T Sigma) C (iter Sigma d) x)
```

The pattern functor

Object and arrow parts

```
F  :: Sig -> Set -> Set
F' :: (Sigma :: Sig) -> (X,Y :: Set)
    -> (X -> Y)
    -> F Sigma X -> F Sigma Y
```

are defined by

$$\begin{aligned} \mathbf{F}_{[n_1, \dots, n_m]} X &= X^{n_1} + \dots + X^{n_m} \\ \mathbf{F}'_{[n_1, \dots, n_m]} XY f &= f^{n_1} + \dots + f^{n_m} \end{aligned}$$

Remark. It is possible to modify the elimination rule to account for generic primitive recursion rather than just iteration.

Generic dependent type theory

It is possible to encode a large class of inductive types (including the T_Σ) using well-orderings, but to derive the rules we need extensional type theory (see Dybjer 1997).

A generic formulation of dependent type theory with inductive-recursive definitions was given by Dybjer and Setzer 1999 and for indexed inductive-recursive definitions by Dybjer and Setzer 2001.

The rules for initial algebras can be derived in this theory. In fact, the Dybjer-Setzer axiomatization is obtained by considering a more general universe of signatures and a modified initial algebra diagram.

Generic size

A special case of the initial algebra diagram. Let $\Sigma = [n_1, \dots, n_m]$.

$$\begin{array}{ccc}
 T_{\Sigma}^{n_1} + \dots + T_{\Sigma}^{n_m} & \xrightarrow{\text{Intro}_{\Sigma}} & T_{\Sigma} \\
 \downarrow \text{size}_{\Sigma}^{n_1} + \dots + \text{size}_{\Sigma}^{n_m} & & \downarrow \text{size}_{\Sigma} \\
 \text{Nat}^{n_1} + \dots + \text{Nat}^{n_m} & \xrightarrow{\text{step}_{\Sigma}} & \text{Nat}
 \end{array}$$

where

$$\text{step}_{\Sigma}(\text{In}_i(x_1, \dots, x_{n_i})) = 1 + x_1 + \dots + x_{n_i}$$

can be defined by induction on Σ .

5. Metaprogramming. Well-typed interpreters and partial evaluators

A well-typed interpreter

Consider a small typed programming language based on combinators. With dependent types we can formalize this object language as a type-indexed family of terms.

```
Ty :: Set
Te :: Ty -> Set
```

Some types

```
NAT  :: Ty
(=>) :: Ty -> Ty -> Ty
```

Some terms

$(@)$ $:: (A, B :: Ty)$
 $\rightarrow Te (A \Rightarrow B) \rightarrow Te A \rightarrow Te B$

K $:: (A, B :: Ty)$
 $\rightarrow Te (A \Rightarrow B \Rightarrow A)$

S $:: (A, B, C :: Ty)$
 $\rightarrow Te ((A \Rightarrow B \Rightarrow C)$
 $\Rightarrow (A \Rightarrow B) \Rightarrow A \Rightarrow C)$

$ZERO$ $:: Te NAT$

$SUCC$ $:: Te (NAT \Rightarrow NAT)$

$ITER$ $:: (C :: Ty)$
 $\rightarrow Te ((C \Rightarrow C) \Rightarrow C \Rightarrow NAT \Rightarrow C)$

We hide type arguments and write eg $f @ c$ rather than $(@) A B f a$.

Semantics = interpretation

Interpretation of types:

$\text{Eval} :: \text{Ty} \rightarrow \text{Set}$

$\text{Eval NAT} = \text{Nat}$

$\text{Eval (A => B)} = \text{Eval A} \rightarrow \text{Eval B}$

Interpretation of terms:

$\text{eval} :: (\text{A} :: \text{Ty}) \rightarrow \text{Te A} \rightarrow \text{Eval A}$

$\text{eval (f @ c)} = \text{eval f (eval c)}$

$\text{eval K} = k = \lambda x y \rightarrow x$

```
eval S      = s = \x y z -> (x z) y z
eval ZERO   = Zero
eval SUCC   = Succ
eval ITER   = iter
```

as usual hiding the type argument. `iter` is the iterator

```
iter :: (C :: Set) -> (C -> C) -> C -> Nat -> C
```

Partial evaluation

Consider the function

```
power :: Nat -> Nat -> Nat
```

```
power m n = iter (mult m) 1 n
```

```
mult m n = iter (add m) 0 n
```

```
add m n = iter Succ m n
```

where we have hidden the first argument Nat of iter.

Static and dynamic arguments

In partial evaluation we distinguish between *binding-times*, that is,

static arguments, which are known, and

dynamic arguments, which are not known

at *specialization time*. If m is dynamic and n is static in `power m n` then we can specialize and simplify the definition. Eg $n = 3$

```
power m 3 = iter (mult m) 1 3
           = mult m (mult m (mult m 1))
           = mult m (mult m m)
```

The simplified program is called the *residual program*.

2-level lambda calculus

In 2-level lambda calculus types and terms are given binding-time annotations. Eg the type `Nat` exists in both a static version $\overline{\text{Nat}}$ and a dynamic version $\underline{\text{Nat}}$.

The function `power` with a first dynamic and a second static argument thus gets the type:

$$\begin{aligned} \text{powerDS} &:: \underline{\text{Nat}} \rightarrow \overline{\text{Nat}} \rightarrow \underline{\text{Nat}} \\ \text{powerDS } m \ n &= \overline{\text{iter}} (\underline{\text{mult}} \ m) (\$ \ 1) \ n \end{aligned}$$

where

$$\$:: \overline{\text{Nat}} \rightarrow \underline{\text{Nat}}$$

transforms a static number into the corresponding dynamic one.

Binding-times

There are four different versions of `power` depending on the binding-times of the arguments:

$$\text{powerDS} :: \underline{\text{Nat}} \rightarrow \overline{\text{Nat}} \rightarrow \underline{\text{Nat}}$$
$$\text{powerSD} :: \overline{\text{Nat}} \rightarrow \underline{\text{Nat}} \rightarrow \underline{\text{Nat}}$$
$$\text{powerDD} :: \underline{\text{Nat}} \rightarrow \underline{\text{Nat}} \rightarrow \underline{\text{Nat}}$$
$$\text{powerSS} :: \overline{\text{Nat}} \rightarrow \overline{\text{Nat}} \rightarrow \overline{\text{Nat}}$$

Binding-times again

With dependent types and the correspondences

dynamic	object language
static	meta language

to get types for “generating extensions” corresponding to the binding time annotations:

```
powerDS :: Nat -> Te (NAT => NAT)
```

```
powerSD :: Nat -> Te (NAT => NAT)
```

```
powerDD :: Te (NAT => NAT => NAT)
```

```
powerSS :: Nat -> Nat -> Nat
```

Note the analogy between the binding-time annotated types and the dependent types, except that `powerDS` exchanges the order of its arguments, because a static argument is always given before a dynamic one.

Terms-in-context

Combinatory logic is not quite suitable for partial evaluation, and we want to work in lambda calculus instead. Therefore we need to formalize terms-in-context. We use a name-free approach with

$$\text{Te} :: [\text{Ty}] \rightarrow \text{Ty} \rightarrow \text{Set}$$

so that $\text{Te } [A_1, \dots, A_n] A$ is the set of terms of type A , where the variables (de Bruijn indices) have the types A_1, \dots, A_n .

Pure typed lambda terms are then generated by the following rules:

$$\begin{aligned} (\text{C}) &:: (\text{As} :: [\text{Ty}]) \rightarrow (A, B :: \text{Ty}) \\ &\rightarrow \text{Te } \text{As } (A \Rightarrow B) \rightarrow \text{Te } \text{As } A \rightarrow \text{Te } \text{As } B \end{aligned}$$

```
LAM :: (As :: [Ty]) -> (A,B :: Ty)
      -> Te (A:As) B -> Te As (A => B)
VAR :: (As :: [Ty]) -> (A :: Ty)
      -> Member A As -> Te As A
```

We hide context and type arguments.

A well-typed interpreter for typed lambda terms

The interpretation of types is as before, but now we need to interpret contexts too:

```
Eval :: [Ty] -> Set
```

```
Eval [] = ()
```

```
Eval (A:As) = (Eval A, Eval As)
```

Interpretation of terms:

```
eval :: (As :: [Ty]) -> (A :: Ty)  
      -> Eval As -> Eval A
```

```
eval (f @ c) as = eval f as (eval c as)
eval (LAM e) as = \x -> eval e (x,as)
eval (VAR n) as = proj n as
```

again hiding the context and type arguments.

Partially evaluating power again

When we specialize `power` with a static second argument `n` we get the term

```
\x -> iter (mult x) 1 n :: Nat -> Nat
```

which we want to simplify by using that `iter` is a static operation.

The corresponding object language term is:

```
LAM (ITER @ (MULT @ (VAR X0)) @ ($ 1) @ ($ n))  
:: Te [] (NAT => NAT)
```

where

```
$ :: (As :: [Ty]) -> Nat -> Te As NAT
```

is the injection of a metalanguage natural number into the object language:

$$\$ \text{ Zero} = \text{ZERO}$$
$$\$ (\text{Succ } n) = \text{SUCC } (\$ n)$$

Some typings

In the term

```
LAM (ITER @ (MULT @ (VAR XO)) @ ($ 1) @ ($ n))  
:: Te [] (NAT => NAT)
```

we use the following instances:

```
ITER :: Te [NAT]  
      ((NAT => NAT) => NAT => NAT => NAT)  
MULT :: Te [NAT] (NAT => NAT => NAT)  
XO   :: Member NAT [NAT]
```

Executing a static operation

The fact that `iter` is a static operation here is expressed by the fact that for all $n :: \text{Nat}$ the object language term obtained by executing

```
iter (\t -> MULT @ (VAR X0) @ t) ($ 1) n
```

has the same semantics as

```
ITER @ (MULT @ (VAR X0)) @ ($ 1) @ ($ n)
```

For $n = 2$:

```
iter (\t -> MULT @ (VAR X0) @ t) ($ 1) 2  
= MULT @ (VAR X0) @ (MULT @ (VAR X0) @ ($ 1))
```

This term can be further simplified to the semantically equal term

```
MULT @ (VAR X0) @ (VAR X0) :: Te [Nat] Nat
```

This is the residual program (normal form).

6. Coping with general recursion

Recursion in type theory

Total type theory. Recursion in Martin-Löf type theory is primitive (structural) recursion. General recursive algorithms are not directly typable in total type theory. This is one of the main obstacles to making total type theory a programming language.

Partial type theory. There is a version of Martin-Löf type theory with general recursion (Martin-Löf's domain interpretation of type theory 1983, Palmgren 1991). Cayenne can be said to be an implementation of partial type theory. Partial type theory allows non-terminating computations and does not support the Curry-Howard correspondence.

We want both logic and general recursion. How? There are many suggestions.

We shall look at a method for generating special purpose accessibility predicates for general recursive definitions (Bove 1999 and Bove and Capretta 2001).

Quicksort in Haskell

```
qSort :: [Nat] -> [Nat]
```

```
qSort [] = []
```

```
qSort (x : xs)
```

```
  = qSort (filter (< x) xs)
```

```
    ++ x : qSort (filter (>= x) xs)
```

(A more efficient version which partitions `xs` in one pass can easily be written.)

Quicksort in total type theory

We can define a termination predicate for quicksort:

```
D :: [Nat] -> Set
C0 :: D []
C1 :: (x :: Nat) -> (xs :: [Nat])
     -> D (filter (< x) xs)
     -> D (filter (>= x) xs)
     -> D (x : xs)
```

Quicksort can then be represented as a function of two arguments: a list and a proof that quicksort terminates for this list.

```
qSort :: (xs :: [Nat]) -> D xs -> [Nat]
```

```
qSort [] C0 = []
```

```
qSort (x : xs) (C1 x xs p q)
```

```
  = qSort (filter (< x) xs) p
```

```
    ++ x : qSort (filter (>= x) xs) q
```

Termination of quicksort

Quicksort terminates for all lists:

```
qSortTerm :: (xs :: [Nat]) -> D xs
```

Hence

```
\xs -> qSort xs (qSortTerm xs)  
:: [Nat] -> [Nat]
```

Treesort

```
treeSort = buildBST . preorder
```

```
buildBST :: [Nat] -> BST
```

```
preorder :: BST -> [Nat]
```

This is essentially the same algorithm as functional quicksort, but it is structurally recursive!

(C. McBride)

The idea of *strong* functional programming!

(D. Turner)

McCarthy's 91-function in Haskell

The Bove-Capretta method is applicable to nested recursion as well.

Haskell code for McCarthy's 91-function:

```
f91 :: Nat -> Nat
```

```
f91 n = if n > 100 then n - 10  
        else f91 (f91 (n + 11))
```

McCarthy's 91-function in total type theory

We get a *simultaneous inductive-recursive definition* of the termination predicate and the structural recursive version of f91:

```
D    :: Nat -> Set
f91  :: (n :: Nat) -> D n -> Nat

C0  :: (n :: Nat) -> T (n > 100) -> D n
C1  :: (n :: Nat) -> T (n <= 100)
      -> (p :: D (n + 11))
      -> D (f91 (n + 11) p)
      -> D n
```

f91 n (C0 n r) = n - 10

f91 n (C1 n r p q)

= f91 (f91 (n + 11) p) q

Summary

- Dependent types for ML/Haskell style programming:
 - classes, modules
 - arrays, sized types, simple datatype invariants
 - generic programming
 - metaprogramming
- Correctness in the short term:
 - modest use of dependent types (eg a la DML)
 - combining testing and dependent types
- General recursion and dependent types.