# Language Engineering (TB1)

## Taken by Joseph MacManus on a course given by Dr Nicolas Wu (2018)

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## 1 DSLs and Catamorphisms

## 1.1 Domain-Specific Languages

A programming language consists of three main components:

- Syntax the shape of grammer / words / vocab.
- Semantics the meaning; a function from syntax to some domain.
- Pragmatics The purpose of a language.

A domain-specific language (DSL) is a language that has been crafted with a specific purpose in mind. These are not necessarily Turing-complete. Some DSLs come equipped with all the features of general purpose languages:

- Parser
- Syntax highlighting
- IDE
- Compiler
- Documentation

and so on. An embedded DSL (EDSL) is defined within another host language. The advantage is that there is less work to perfom, but this is at the cost of restricted flexibility. An EDSL can be either a *deep* or *shallow* embedding. A deep embedding is where the syntax is given by concrete datatypes, and the semantics given by evaluation. A shallow embedding has syntax borrowed directly from it's host language, and semantics is directly given.

For example, consider:

"
$$3+5$$
" in a string. [ $3+5$ ] :: Int

Where [] are denotational brackets. We use them to ascribe a semantics.

$$[3+5] = [3] + [5] = 3+5=8$$

We can model this using a deep embedding in Haskell with the following code:

So instead of [3+5], we can now write eval (Add (Var 3) (Var 5)).

The shallow embedding is given directly by functions: (We will redefine Expr here)

```
type Expr = Int

var :: Int \rightarrow Expr

var n = n

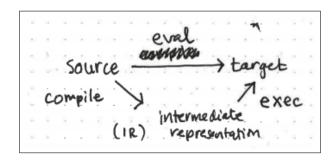
add :: Expr \rightarrow Expr \rightarrow Expr

add x y = x + y
```

Our example is now written as:

```
add (var 3) (var 5)
```

*Remark.* What is a compiler, then? A compiler is code that transforms a language to a target language through some intermediate representation.

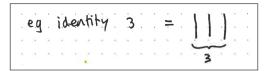


Typical examples of this diagram include the C compiler gcc, which takes a C source file and compiles this to assembly. That assembly is then executed in the terminal. Javascript tends to ignore the compile stage since it is an interpreted language. The web browser performs eval, which turns JS into rendered output. Haskell has two modes, when using GHC, it compiles .hs files into assembly, which can then be executed in the terminal. However, when using GHCi, it takes source code and interprets it directly by evaluating in the terminal.

## 1.2 Case Study: Circuit Language

We will study a particular example of a DSL, and the different ways to embed it in Haskell. The circuit language is a DSL for describing circuits. It consists of several operations. Rather than define each operation formally, we shall give the type of each operation and an example of what it does.

• identity :: Int  $\rightarrow$  Circuit

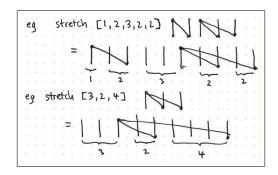


• beside :: Circuit  $\rightarrow$  Circuit  $\rightarrow$  Circuit

• above :: Circuit → Circuit → Circuit

• fan :: Int → Circuit

• stretch :: [Int]  $\rightarrow$  Circuit  $\rightarrow$  Circuit



This language is used to describe how circuits work, information starts at the top of each line and travels downwards. Information is combined by joining lines, and applying the associative operation of our choosing:

eg. 
$$x_0 \times x_1 \times x_2$$
  $y = y_0 \cdot x_1 \cdot x_2$ 

And information is duplicated when lines seperate:

eg 
$$x = y_0 = y_1 = y_2$$

There are many different ways of interpeting this circuit language. For example, we may want to simply find the width or height of a circuit, or perhaps we will evaluate the output

of a circuit given an input and the node operation. We start with a deep embedding. This is achieved in two steps:

- 1. Create a data structure for the abstract syntax.
- 2. Define a semantics with an evaluation function.

The first step isn't too hard, we simply get:

Let's interpet the circuit language by stipulating that the semantics of a term is the width of the circuit generated. We will define a semantics with a function width.

We can have multiple semantics easily by supplying more evaluation functions. For instance, the height of the circuit is:

```
type Height = Int \begin{array}{l} \text{height} \ :: \ \text{Circuit} \ \to \ \text{Height} \\ \text{height} \ (\text{Identity n}) \ = \ 1 \\ \dots \\ \text{height} \ (\text{Above} \ c_1 \ c_2) \ = \ \text{height} \ c_1 \ + \ \text{height} \ c_2 \end{array}
```

Sometimes the semantics are interwined in a dependent way. For instance, calculating if a circuit is well connected requires us to calculate the width even though all we are interested in is one bool.

In a shallow embedding we simply have to give a semantics in terms of the operations directly.

```
type Width = Int

type Circuit = Width

identity :: Int \rightarrow Circuit

identity n = n

beside :: Circuit \rightarrow Circuit \rightarrow Circuit

beside c_1 c_2 = c_1 + c_2

above :: Circuit \rightarrow Circuit \rightarrow Circuit

above c_1 c_2 = c_1

fan :: Int \rightarrow Circuit

fan n = n

stretch :: [Int] \rightarrow Circuit \rightarrow Circuit

stretch ws c = sum ws
```

Shallow is problemaic because it is hard to add a different semantics. In order to do so we must redefine Circuit, but this might break any code that depends on the old definition. Additionally, a dependent semantics requires all of the interpretations to be considered at once. This is not compositioned. However, in a shallow semantics it is easy to extend the language with new operations, since this involves adding new functions: nothing breaks. With a deep embedding a new constructor must be added, so all semtantics must be extended accordingly.

*Is it possible to extend the syntax and semtnaics of a language in a modular fashion?* This is known as *The Expression Problem.* For instance, consider the data type Expr from before:

```
data Expr =val Int | Add Expr Expr
```

Here we want to extend the syntax by adding a new operation for multiplication, but we do not want to modify any existing code. I.e. we cannot simply add a new Mul constructor. Similarly, consider the semantics:

```
eval :: Expr \rightarrow Int
```

Again, we want to extend the semantics without modifying the code. (Though in this case adding semantics is easy as we simply write another function of type  $Expr \rightarrow b$ ). To solve the expression problem, we will study a generalisation of folds called a *catamorphism*. We do this because folds are a way of reducing data structures in a composable way, and syntax trees are just data structures.

## 1.3 Catamorphisms

Consider the fold for a list:

```
data [a] = []

| a : [a]

foldr :: b \rightarrow (a \rightarrow b \rightarrow b) \rightarrow [a] \rightarrow b

foldr k f [] = k

foldr k f (x:xs) = f x (foldr k f xs)
```

How do we generalise this? Our first step is to deconstruct the type of lists to expose its generic structure. The defintion of lists is the same as the following, when we remove the syntactic sugar.

We remove recursion from this data type, and mark it with a parameter k, for *k*ontinuation.

Next, we make the recursive paramter something we can change programatically by giving a Functor instance to List a

```
instance Functor (\underline{List} a) where fmap :: (x \rightarrow y) \rightarrow \underline{List} a x \rightarrow \underline{List} a y fmap f \underline{Empty} = \underline{Empty} fmap (Cons a x) = Cons a (f x)
```

We now need to derive a type that gives us the fixed point of data. This is defined as follows:

```
data Fix f = In (f (Fix f))
```

This datatype allows us to generalise all recursive data types (except mutually recursive ones). For example, instead of List a, we can write Fix (<u>List</u> a). To demonstrate this, we show that List a and Fix (<u>List</u> a) are isomorphic.

```
toList :: Fix (\underline{List} a) \rightarrow List a fromList :: List a \rightarrow Fix (List a)
```

We say that List a and Fix ( $\underline{\text{List}}$  a) are isomorphic when composing the above functions in any order yields the identity. Let's define these functions.

```
fromList :: List a \rightarrow Fix (<u>List</u> a) fromList Empty = In <u>Empty</u>
```

Some examples of values of type Fix (List a) are:

```
In Empty :: Fix (List a)
```

(Note that the type of In is In :: f (Fix f)  $\rightarrow$  Fix f so the example above is where f is List a.)

```
In (Cons 5 (In Empty))
```

and for two elements we have

```
In (Cons 6 (In (Cons 7 (In Empty))))
```

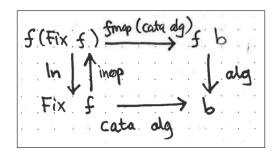
So now we have enough to finish a definition of fromList.

```
fromList :: List a \rightarrow Fix (<u>List</u> a)
fromList Empty = In <u>Empty</u>
fromList (Cons x xs) = In (<u>Cons</u> x (fromList xs))
```

We are now ready to generise fold to be a catamorphism. Consider functions of the form List a  $b \rightarrow b$ :

```
\begin{array}{lll} h :: \underline{List} & a & b & \rightarrow & b \\ h & \underline{Empty} & = & k \\ h & (\underline{Cons} & a & y) & = & f & a & y \\ & & where & & \\ & & k & :: & b \\ & & f & :: & a & \rightarrow & b & \rightarrow & b \end{array}
```

Functions of that type correspond to replacing the constructors of <u>List</u> a with functions k and f just like in foldr. A catamorphism arises from this diagram:



The function inop is the opposite of In. We define it by the following:

```
inop :: Fix f \rightarrow f (Fix f) inop (In x) = x
```

So finally we can write the function cata by chasing the arrows of this square:

```
cata :: Functor f \Rightarrow (f b \rightarrow b) \rightarrow Fix f \rightarrow b cata alg x = (alg \circ fmap (cata alg) \circ inop) x
```

An alternative and equivalent definition is:

```
cata alg (In x) = alg (fmap (cata alg) x)
```

To use this, we only need to supply the alg. We will define the function toList using a cata:

```
toList :: Fix (\underline{List} \ a) \rightarrow List \ a toList = cata alg where alg :: \underline{List} \ a \ (List \ a) \rightarrow List \ a alg \underline{Empty} = Empty alg Cons \ x \ xs = Cons \ x \ xs
```

#### or, equivalently

```
toList' :: Fix (\underline{List} \ a) \rightarrow [a]
toList' = cata alg
where
alg :: \underline{List} \ a \ [a] \rightarrow [a]
alg \underline{Empty} = []
alg \underline{Cons} \ x \ xs = x : xs
```

We can also define a function that returns the length of a Fix (List a).

```
length :: Fix (\underline{List} \ a) \rightarrow Int
length :: cata alg
where
alg :: \underline{List} \ a \ Int \rightarrow Int
alg \underline{Empty} = 0
alg (\underline{Cons} \ x \ y) = y + 1
```

Here is an example of evaluation.

```
length (In (Cons 7 (In (Cons 9 (In Empty)))))
= {length}
  cata alg (In (Cons 7 (In (Cons 9 (In Empty)))))
= {cata}
  alg (fmap (cata alg) (Cons 7 (In ...)))
= \{fmap\}
  alg (Cons 7 (cata alg (In (Cons 9 (In Empty)))))
= \{alg\}
  1 + cata alg (In (Cons 9 (In Empty)))
= {cata}
  1 + alg (fmap (cata alg) (Cons 9 (In Empty)))
= \{fmap\}
  1 + alg (Cons 9 (cata alg (In Empty)))
= \{alg\}
  1 + 1 + cata alg (In Empty)
= {cata}
  1 + 1 + alg (fmap (cata alg) Empty)
= \{fmap\}
  1 + 1 + alg Empty
= \{alg\}
  1 + 1 + 0
= \{(+)\}
  2
```

Now another exmaple, of summing a list.

```
\begin{array}{l} \text{sum} \ :: \ \text{Fix} \ (\underline{\text{List}} \ \text{Int}) \ \rightarrow \ \text{Int} \\ \\ \text{sum} \ = \ \text{cata alg} \\ \\ \text{where} \\ \\ \text{alg} \ :: \ \underline{\text{List}} \ \text{Int} \ \text{Int} \ \rightarrow \ \text{Int} \\ \\ \text{alg} \ \underline{\text{Empty}} \ = \ 0 \\ \\ \text{alg} \ (\text{Cons} \ x \ y) \ = \ x \ + \ y \end{array}
```

### 1.4 Case Study: Peano Numbers

A Peano number is either zero, or a successor of another Peano number.

So the number 3 is written S(S(S(0))). Our goal is to create a modular DSL which models Peano numbers, a fashion similar to above. Step 1: make a signature functor for Peano.

```
data \underline{Peano} k = z | S k
```

Step 2: Define a functor instance for Peano.

```
instance Functor \underline{Peano} where fmap :: (a \rightarrow b) \rightarrow \underline{Peano} a \rightarrow \underline{Peano} b fmap f z = z fmap f (S x) = S (f x)
```

Step 3: Write functions using cata.

```
toInt :: Fix \underline{Peano} \rightarrow Int
toInt = cata alg where
alg :: \underline{Peano} Int \rightarrow Int
alg z = 0
alg (S x) = x + 1
```

Now we can define a doubling function.

```
\begin{array}{lll} \mbox{double} & :: \mbox{ Fix } \mbox{ \underline{Peano}} & \rightarrow \mbox{ Fix } \mbox{ \underline{Peano}} \\ \mbox{double} & = \mbox{cata alg where} \\ & \mbox{alg } :: \mbox{ \underline{Peano}} \mbox{ (Fix } \mbox{ \underline{Peano}}) & \rightarrow \mbox{ Fix } \mbox{ \underline{Peano}} \\ \mbox{alg } z & = \mbox{ In } z \\ \mbox{alg } (\mbox{S} \mbox{ x}) & = \mbox{ In } (\mbox{S} \mbox{ (In } (\mbox{S} \mbox{ x}))) \end{array}
```

#### 1.5 Composing Languages

Previously, we looked at the following data type as the language for addition.

We then learnt to extract the signature functor for this by locating recursive calls:

The Fix  $\underline{\mathtt{Expr}}$  datatype is essentially  $\underline{\mathtt{Expr}}$ . Suppose we want to add multiplication to this language. We need a way to extend  $\underline{\mathtt{Expr}}$  with more constructors. This is achieved by the coproduct of functors. The coproduct functor is defined as:

```
data (f :+: g) a = L (f a)
| R (g a)
```

This takes two functors and makes the functor (f:+:g). It introduces these constructors:

```
L :: f a \rightarrow (f:+:g) a
R :: g a \rightarrow (f:+:g) a
```

The functor instance is defined as follows

```
instance (Functor f, Functor g) \Rightarrow Functor (f:+:g) where fmap :: (a \rightarrow b) \rightarrow (f:+:g) a \rightarrow (f:+:g) b fmap f (L x) = L (fmap f x) fmap f (R y) = R (fmap f y)
```

Now we are ready to define the signature functor for multiplication:

```
data \ \underline{Mul} \ k = \underline{Mul} \ k \ k
```

This is the datatype constructor:

```
\texttt{Mul} \; :: \; k \; \rightarrow \; k \; \rightarrow \; \texttt{Mul} \; \; k
```

We define its functor instance:

```
instance Functor \underline{\text{Mul}} where fmap f (Mul x y) = Mul (f x) (f y)
```

Finally, we can put the  $\underline{\mathtt{Expr}}$  and  $\underline{\mathtt{Mul}}$  languages together, to have a language with both addition and multiplication.

```
Fix (Expr :+: Mul)
```

This is essentially the same as describing the following datatype, but in a compositional way:

For practical purposes, we do not work with Expr' but with Fix ( $\underline{\mathtt{Expr}}:+:\underline{\mathtt{Mul}}$ ). We need to write algebras<sup>1</sup> of the form:

```
Expr :+: Mul b \rightarrow b
```

to reduce a Fix (<u>Expr</u>:+: <u>Mul</u>) type to b. To do this in a compositional way, we define a way of combining Expr algebras and Mul algebras. We call this the junction of algebras:

```
(\nabla) :: (f a \rightarrow a) \rightarrow (g a \rightarrow a) \rightarrow ((f:+:g) a \rightarrow a) (falg \nabla galg) (L x) = falg x (falg \nabla galg) (R y) = galg y
```

So now, we can give a semantics to the language Fix (Expr: +:Mul) by defining algebras.

<sup>&</sup>lt;sup>1</sup>An *algebra* is any function of type  $f a \rightarrow a$ , where f is a functor.

```
add :: \underline{Expr} Int \rightarrow Int
add (\underline{Val} x) = x
add (\underline{Add} x y) = x + y
mul :: \underline{Mul} Int \rightarrow Int
mul (\underline{Mul} x y) = x * y
```

To evaluate, we write:

```
eval :: Fix (\underline{Expr} : +: \underline{Mul}) \rightarrow Int eval = cata (add \nabla mul)
```

And thus, we have solved the expression problem. In fact, we can decompose the  $\underline{\mathtt{Expr}}$  into constituent parts:

```
data \underline{\text{Val}} k = \underline{\text{Val}} Int data \underline{\text{Add}} k = \underline{\text{Add}} k k
```

After defining functor instances, we can define functor instances for:

```
Fix (Val :+: Add :+: Mul)
```

#### 2 Grammars and Parsers

#### 2.1 BNF: Baccus-Naur Form

BNF is a language used to express the shape of grammars. It was invented in around 1958 for the expression of the Algol programming language. A BNF statement is made up of:

- $\varepsilon$  represents empty strings.
- <n> represents a non-terminal.
- "x" represents a terminal.
- p|q represents a choice between p and q.

An example of BNF is the following:

We can approximate numbers by this:

The core language of BNF is usually extended with some constructs:

```
[ e ] // optional e
( e ) // grouping e
  e* // 0 or more repetitions of e
  e+ // 1 or more repetitions of e
```

For a more complex example, consider expressions:

This corresponds to the following type:

```
data Expr = Val Num
| Add Num Expr
```

In principle, we do the same to convert <num> into a Num datatype. However, we will approximate this by the type Int.

```
type Num = Int
```

Grammars can sometimes be ambiguous, a single string can be accepted by the grammar in multiple ways:

The problem here is also that <expr> is left-recursive: there is a branch which has an <expr> before any terminal. This causes problems in recursive descent parsers, which we will study later. The solution is to refactor the grammar.

## 2.2 Paull's Modified Algorithm

We can remove recursion as follows. Suppose we have the following grammar:

```
\mathtt{A} \; ::= \; \mathtt{A} \;\; \alpha_1 \;\; | \;\; \cdots \;\; | \;\; \mathtt{A} \;\; \alpha_n \;\; | \;\; \beta_1 \;\; | \;\; \cdots \;\; | \;\; \beta_m
```

To apply the algorithm, we rewrite the term above to be the following:

In practive here is how we convert the following:

#### 2.3 Parsers

A parser is a function that takes in a list of characters, and returns an item that was parsed, along with the unconsumed string. We can define it by:

```
newtype Parser a = Parser (String \rightarrow [(a, String)])
```

We can use a parser by defining a function, parse.

```
parse :: Parser a → String → [(a, String)]
parse (Parser px) = px

(px :: String → [(a, String)])
```

Some very simple examples of parsers include the fail parser which always fails, the item parser which knows how to process a single character, and the look parser, which allows one to look into the input stream without consuming anything.

Often we want to transform our parsers from producing values of one type to another. For instance, we might transform a parser for a single Char into producing the corresponding Int. This is achieved by giving a functor instance for parsers. We can use this to define a Parser for Ints from our item parser (see the homework).

```
instance Functor Parser where

-- fmap :: (a \rightarrow b) \rightarrow Parser \ a \rightarrow Parser \ b

fmap f (Parser px) =

Parser ( \lambda ts \rightarrow [(f \ x, \ ts') \mid (x, \ ts') \leftarrow px \ ts])

(\langle \$ \rangle) :: (a \rightarrow b) \rightarrow Parser \ a \rightarrow Parser \ b

f \langle \$ \rangle px = fmap f px
```

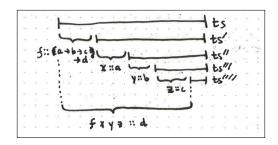
The following variation is often useful:

```
(\langle \$ \rangle :: a \rightarrow Parser b \rightarrow Parser a x \langle \$ py = fmap (const x) py
```

We can use this to build a function called skip that parses input but outputs nothing useful.

```
skip :: Parser a \rightarrow Parser ()
skip px = () \langle \$ px \rangle
```

Now we want to apply a function to the different outputs of parse:



To make something like this we use a combination of  $\langle * \rangle$  and  $\langle * \rangle$  like this:

```
pf \langle * \rangle px \langle * \rangle py \langle * \rangle pz
```

This uses the  $(\langle * \rangle)$  operation, which we will define shortly. The applicative introduces pure and  $(\langle * \rangle)$ .

```
class Functor f \Rightarrow Applicative f where pure :: a \rightarrow f a (\langle * \rangle) :: f (a \rightarrow b) \rightarrow f a \rightarrow f b
```

These have the following definitions:

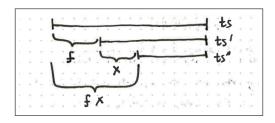
```
instance Application Parser where
-- pure :: a \rightarrow Parser a
pure x = Parser (\lambda ts \rightarrow [(x,ts)])
```

The pure x parser will not consume any input but always generates the value x.

```
parse (pure 5) "hello" = [(5, "hello")]
```

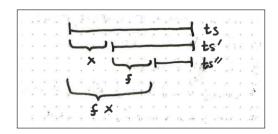
Now we define  $(\langle * \rangle)$ , pronounced "ap", for apply.

```
-- (\langle * \rangle) :: Parser (a \rightarrow b) \rightarrow Parser a \rightarrow Parser b
Parser pf \langle * \rangle Parser px = Parser ( \lambdats \rightarrow [(f x, ts'') \leftarrow pf ts , (x, ts'') \leftarrow px ts'])
```



The operation we defined first parses a function, then a value, and finally applies the function to the value. This can be done the other way around too:

```
\begin{array}{lll} (\langle **\rangle) & :: \ \mathsf{Parser} \ \mathsf{a} \ \to \ \mathsf{Parser} \ \mathsf{(a} \ \to \ \mathsf{b)} \ \to \ \mathsf{Parser} \ \mathsf{b} \\ \mathsf{Parser} \ \mathsf{px} \ \langle **\rangle \ \mathsf{Parser} \ \mathsf{pf} \ = \ \mathsf{Parser} \ (\ \lambda \mathsf{ts} \ \to \\ & \ [ \ (\mathsf{f} \ \mathsf{x}, \ \mathsf{ts'})') \\ & \ | \ (\mathsf{x}, \ \mathsf{ts'})' \ \leftarrow \ \mathsf{px} \ \mathsf{ts} \\ & \ , \ (\mathsf{f}, \ \mathsf{ts'})' \ \leftarrow \ \mathsf{pf} \ \mathsf{ts'} \ ]) \end{array}
```



Other derived operators are  $(\langle * \rangle)$  and  $(* \rangle)$ , their types are:

```
(\langle * \rangle :: Parser a \rightarrow Parser b \rightarrow Parser a (* \rangle) :: Parser a \rightarrow Parser b \rightarrow Parser b
```

#### 2.4 Monoidal

The Monoidal class is equivalent to the Applicative class:

For parsers, the monoidal instance is defined as follows:

```
instance Monoidal Parser where

-- unit :: Parser ()

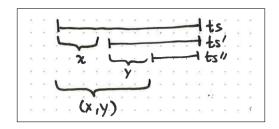
unit = Parser ( \lambdats \rightarrow [((), ts)])

-- mult :: Parser a \rightarrow Parser b \rightarrow Parser (a,b)

mult (Parser px) (Parser py) =

Parser (\lambdats \rightarrow [ ((x,y), ts'')

| (x, ts') \leftarrow px ts
|, (y, ts'') \leftarrow py ts'])
```



It is useful to make mult a binary operation, so we introduce one:

```
(\langle \sim \rangle) :: Monoidal f \Rightarrow f a \rightarrow f b \rightarrow f (a,b) px \langle \sim \rangle py = mult px py
```

We the derive these useful combinators:

```
(\langle \sim \rangle) :: Monoidal f \Rightarrow f a \rightarrow f b \rightarrow f a
px \langle \sim py = fst \langle \$ \rangle \ (px \langle \sim \rangle \ py)
(\sim) :: Monoidal f \Rightarrow f a \rightarrow f b \rightarrow f b
px \sim \rangle \ py = snd \langle \$ \rangle \ (px \langle \sim \rangle \ py)
-- where fst \ (x,y) = x, snd \ (x,y) = y.
```

Note that we have this equivalence:

- $(\langle \sim ) = (\langle * )$
- $(\sim)) = (*)$

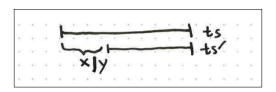
#### 2.5 Alternatives

Now we produce parsers that can deal with choice in a grammar.

```
class Alternative f where empty :: f a  (\langle | \rangle) \ :: \ f \ a \ \rightarrow \ f \ a \ \rightarrow \ f \ a
```

Here is the instance for parsers:

```
instance Alternative Parser where
-- empty :: Parser a
empty = fail -- from before
-- (\langle | \rangle) :: Parser a \rightarrow Parser a \rightarrow Parser a
Parser px \langle | \rangle Parser py = Parser (\lambdats \rightarrow px ts ++ py ts )
```



```
-- parse px ts ++ parse py ts = parse (px \langle | \rangle py) ts
```

Sometimes we want to extend  $\langle | \rangle$  to many input parsers.

```
choice :: [Parser a] \rightarrow Parser a choice pxs = foldr (\langle | \rangle) empty pxs
```

We can define a combinator that appends the result of a parse onto others:

```
(\langle : \rangle) :: Parser a \rightarrow Parser [a] \rightarrow Parser [a] px \langle : \rangle pxs = (:) \langle \$ \rangle px \langle * \rangle pxs
```

To understand this, first recall that most parser combinators are left associative.

```
(:) ⟨$\ px ⟨*\ pxs
=

((:) ⟨$\ px) ⟨*\ pxs

-- check the types to see this is correct, recall:
-- (⟨$\) :: (a → b) → f a → f b
-- (⟨*\) :: f (a → b) → f a → f b
```

Now we're ready to define combinators that correspond to + and \* from BNF.

- *e*+ is written some e
- *e*\* is written many e

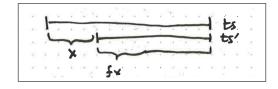
```
some :: Parser a \rightarrow Parser [a] some px = px \langle : \rangle many px
```

this parses one px and appends it to the result of many px

```
many :: Parser a \rightarrow Parser [a] many px = some px \langle | \rangle empty
```

## 2.6 Monadic Parsing

Sometimes we want the control flow a parser to depend on what was parsed. Suppose we have px :: Parser a, we can define a function  $f :: a \rightarrow Parser b$ . The function f inspects the value x which came from px and produces a new parser accordingly. The result should be a parser of time Parser b.



To use this combinator we combine a parser with a function. The satisfy parser takes in a function that is a predicate on Chars and returns the parsed value if it satisfies the predicate.

```
satisfy :: (Char \rightarrow Bool) \rightarrow Parser Char satisfy p = item \gg= \lambda t \rightarrow if p t then pure t else empty
```

This is perhaps the most useful combinator. Rather than the monadic definition, we can write one directly:

```
satisfy :: (Char \rightarrow Bool) \rightarrow Parser Char satisfy p = Parser (\lambdats \rightarrow case ts of [] \rightarrow [] (t:ts') \rightarrow [(t,ts') | p t])
```

We can now parse a single character as follows:

```
char :: Char \rightarrow Parser Char char c = satisfy (c ==)

-- or char c = satisy (\lambdac' \rightarrow c == c')
```

Example:

```
parse (char 'x') "xyz" = [('x',"yz")]
parse (char 'a') "xyz" = []
```

#### 2.7 Chain for left-recursion

The problem with ambiguous grammars that are left-recursive can be resolved with Paull's algorithm.

```
<expr> ::= <number> | <expr> "+" <expr>
```

However, without applying Paull's algorithm, we have a nice data type:

```
\mathtt{data}\ \mathtt{Expr}\ =\ \mathtt{Num}\ \mathtt{Int}\ \mid\ \mathtt{Add}\ \mathtt{Expr}\ \mathtt{Expr}
```

We can decide to use chain11 to parse into this data structure from the original grammar, assuming that + is left associative. (chainr exists if we want it to be right associative). Essentially, we have this combinator:

```
chainl1 :: Parser a \rightarrow Parser (a \rightarrow a \rightarrow a) \rightarrow Parser a
```

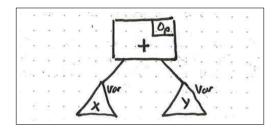
This allows us to write a parser of the form:

```
expr :: Parser Expr expr = chain11 number add add :: Parser (Expr \rightarrow Expr \rightarrow Expr) add = Add \langle$ tok "+"
```

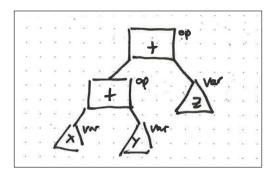
## 3 Abstraction and Semantics

#### 3.1 The Free Monad

Suppose we are interested in giving a semantics to a language for addition. The syntax for this language could look like x+y. This corresponds to a syntax tree:



Or for a more complex example, consider (x+y)+z:



We want to give the shape of + nodes by using a signature functor:

```
data Add k = Add k k
```

In Haskell we can also write:

```
data Add k = k :+ k
```

The provision of variables is left to the Free Monad. The free monad Free f a provides syntax trees whose nodes are shaped by f, and whose variables come from the type a.

It is worth comparing this to the definition of Fix:

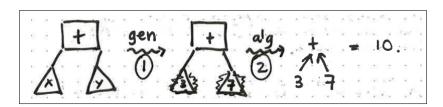
```
data Fix f = In (f (Fix f))
```

The previous trees can be expressed with the following values of type Free Add String.

To interpret these free trees, we work in two stages:

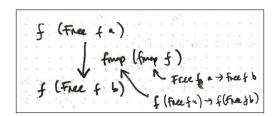
- 1. We change the variables into values. *(the Generator)*
- 2. We evaluate the operations. (*The Algerba*)

In pictures, we do this to interpret a tree<sup>2</sup>:



(*Stage 1*) The first stage involves replacing variables with their corresponding numbers. This is achieved by defining Free f to be a functor. This is only possible is f is a functor too.

```
instance Functor f ⇒ Functor (Free f) where
-- fmap :: (a→b) → Free f a → Free f b
  fmap h (Var x) = Var (h x)
  fmap h (Op op) = Op (fmap (fmap h) op)
-- note that op :: f (Free f a)
```



(Stage 2) The second stage extracts semantics by applying an algebra. This is a recursive function defined as follows (We could use a cata, but that is out of the scope of this lecture series).

```
extract :: Functor f \Rightarrow (f \ b \rightarrow b) \rightarrow Free \ f \ b \rightarrow b extract alg (Var x) = x extract alg (Op op) = alg (fmap (extract alg) op)

-- x :: b, op :: f (Free f b)
```

Finally, we can combine these two stages to define an evaluation function:

```
eval :: Functor f \Rightarrow (f \ b \rightarrow b) \ (a \rightarrow b) \rightarrow Free f \ a \rightarrow b eval alg gen = extract alg \circ fmap gen -- fmap gen is stage 1, extract alg is stage 2
```

<sup>&</sup>lt;sup>2</sup>The triangles on the second tree are an error.

In pictures, we can represent an operation with a box, and a variable with a triangle, and alg will replace boxes, gen will replace triangles. First we define an algebra for a functor. Consider the Add functor from before.

```
add :: Add Int \rightarrow Int add (x :+ y) = x + y
```

We also need a generator from the type of our variables. Variables are often given as strings:

```
type Var = String
```

The Generator for addition is a function from Var to Int.

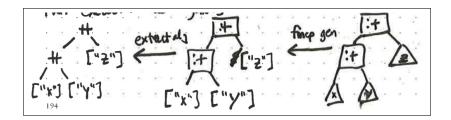
```
env :: Var \rightarrow Int

env "x" = 3

env "y" = 5

env _ = 0
```

This is the environment for evaluation. Suppose we want to evaluate this tree:

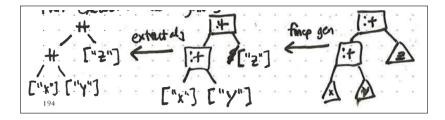


This is an example of eval add env. A second example is to collect all variables in an expression as a list. To do this we provide a function Vars, which is defined used eval.

```
vars :: Free Add Var → [Var]
vars = eval alg gen
where
    gen :: Var → [Var]
    gen x = [x]

alg :: Add [Var] → [Var]
alg (xs :+ ys) = xs ++ ys
```

This executes as follows:



Suppose we want to add an operation to our language, that performs division.

```
data \underline{\text{Div}} k = \underline{\text{Div}} k k
```

If we want to provide a semantics that collects all the variables, we must provide an algebra.

```
divVars :: \underline{\text{Div}} [Var] \rightarrow [Var] divVars (Div xs ys) = xs ++ ys
```

If we want a language with both addition and division, we need to take the coproduct of  $\underline{\mathtt{Add}}$  and  $\underline{\mathtt{Div}}$ . This means expressions of the form  $\underline{\mathtt{Add}}:+:\underline{\mathtt{Div}}$ . For example, we can work with  $\underline{\mathtt{Div}}$  alone:

```
evalDiv :: Free Div Var → [Var]
evalDiv = eval alg gen where
gen :: Var → [Var]
gen x = [x]

alg :: <u>Div</u> [Var] → [Var]
alg (<u>Div</u> xs ys) = xs ++ ys
```

Dealing with both Add and Div at once requires this:

```
vars :: Free (\underline{Add} :+: \underline{Div}) Var \rightarrow [Var] vars = eval alg gen where gen x = [x]

alg :: (\underline{Add} :+: \underline{Div}) [Var] \rightarrow [Var] alg (L (\underline{Add} xs ys)) = xs ++ ys alg (R (\underline{Div} xs ys)) = xs ++ ys
```

When we try to evaluate this language, naively, we encounter a problem.

```
expr :: Free (\underline{Add}:+:\underline{Div}) Var \to Double expr = eval alg gen where gen :: Var \to Double gen = env -- this function magically knows how to assign values to variables alg :: \underline{Add} :+: \underline{Div} \ Double \to \ Double alg (L (\underline{Add} \ x \ y)) = x + y alg (R (\underline{Div} \ x \ y)) = x / y
```

The sad truth is that this function is broken. To see why, consider alg (R ( $\underline{\texttt{Div}}$  x 0)). This will fail! To fix this problem, we must be upfront about the fact that an error can happen. The basic way to do this is to interpret into a Maybe datatype.

```
expr :: Free (\underline{Add} :+: \underline{Div}) Var \rightarrow Maybe Double expr = eval alg gen where gen = env  alg \ (L \ (\underline{Add} \ x \ y)) = mAdd \ -- \ \{note \ x \ :: \ Maybe \ Double\}   alg \ (R \ (\underline{Div} \ x \ y)) = mDiv
```

We must now define mAdd and mDiv, with division we are sensitive to zero:

```
mAdd :: Maybe Double \rightarrow Maybe Double \rightarrow Maybe Double
```

```
mAdd (Just x) (Just y) = Just (x + y)
mAdd mx my = Nothing

mDiv :: Maybe Double → Maybe Double → Maybe Double
mDiv (Just x) (Just 0) = Nothing
mDiv (Just x) (Just y) = Just (x / y)
mDiv mx my = Nothing
```

#### 3.2 Failure

We need to create syntax for failure:

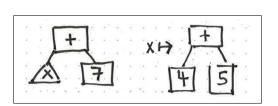
```
data \underline{Fail} k = \underline{Fail} instance Functor \underline{Fail} where \underline{fmap} f \underline{Fail} = \underline{Fail}
```

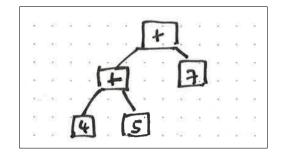
The functor instance shows us that computations can not follow a fail. If we deal with division alone, we have this:

```
evalFail :: Free \underline{\text{Div}} Double \to Free \underline{\text{Fail}} Double evalFail = eval alg gen where gen :: Double \to Free \underline{\text{Fail}} Double gen x = Var x \text{alg :: }\underline{\text{Div}} \text{ (Free }\underline{\text{Fail}} \text{ Double)} \to \text{Free }\underline{\text{Fail}} \text{ Double} alg (\underline{\text{Div}} \text{ (Var x) (Var 0)}) = \text{Op }\underline{\text{Fail}} alg (\underline{\text{Div}} \text{ (Var x) (Var y)}) = \text{Var (x / y)} alg (\underline{\text{Div}} \text{ tl} \text{ tr )} = \text{Op }\underline{\text{Fail}}
```

#### 3.3 Substitution

Substitution in a language is a very useful feature. For example, consider x+7. We can evaluate this into a new syntax tree when we have a notion of substitution, where we might bind x to another expression rather than just a constant, like  $x \to x+5$ , then we expect the above to become (4+5)+7. We can depict this by the following trees.





We will define substitution using code. Usuaully an expression e with a variable x is substituted with e, with the following syntax:

$$e [x \rightarrow e']$$

where e corresonds to x+7, and e' is 4+5. Sometimes we write:

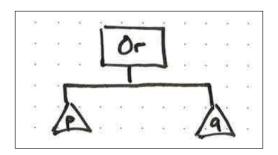
```
e [x \ 4+5]
-- or
e [4+5 / x]
```

For our purposes, a syntax tree is given by a datatype Free f a, where f is the shape of the syantx, and a is the type of the variables, substitution is defined by (>=) as follows:

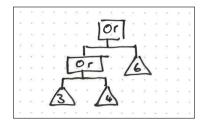
```
(>=) :: Free f a \rightarrow (a \rightarrow Free f b) \rightarrow Free f b
Var x >= f = f x
Op op >= f = Op (fmap (>= f) op)
```

#### 3.4 Non-Determinism

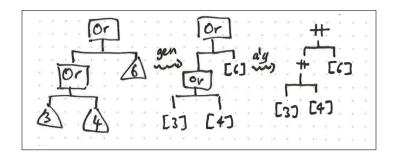
A non-deterministic computation is one that provides the choice between two different computations. For example,  $p \square q$  is the program that gives answers from p or q.



Here we use Or to represent  $\Box$ .



One interprets this tree as follows:



In terms of code, we first need to express syntax, we must also create a functor instance. With this in place, we can define an evaluation function:

```
data Or k = Or k k
instance Functor Or where
   fmap f (Or x y) = Or (f x) (f y)
list :: Free Or a → [a]
list = eval alg gen where
   gen :: a → [a]
   gen :: x = [x]

alg :: Or [a] → [a]
   alg (Or xs ys) = xs ++ ys
```

Another interpretation of these trees is to simply return the first result. We can define the semantics using once.

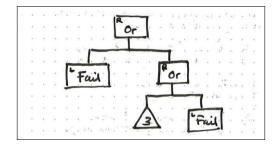
```
once :: Free Or a → Maybe a
once = eval alg gen where
  gen :: a → Maybe a
  gen x = Just x

alg :: Or (Maybe a) → Maybe a
  alg (Or Nothing y) = y
  alg (Or (Just x) y) = Just x
```

This works, but we want a way of signalling that there was no solution. For this, we will make use of Fail. Nondeterminism is the syntax provided by the following synonym.

```
type Nondet a = (Fail :+: Or) a
```

Trees of type Free Nondet a have this shape:



As before, we give semantics to Nondet languages by providing a generator and an algebra.

```
list :: Free Nondet a \rightarrow [a]

list = eval alg gen where

gen :: a \rightarrow [a]

gen x = [a]

alg :: Nondet [a] \rightarrow [a]

alg (L Fail) = []

alg (R (Or x y)) = x ++ y
```

The semantics for once is similar.

#### 3.5 Alternation

An alternative way to do Or is to model a pair of k values as a function from Bool. For this, we define the following:

```
data Alt k = Alt \ (Bool \rightarrow k)
instance Functor Alt where
fmap :: (a \rightarrow b) \rightarrow Alt \ a \rightarrow Alt \ b
fmap f (Alt \ k) = Alt \ (f \circ k)
```

The idea is that we parse True when we want the first child, and False for the second child. Now we can give different semantics for nondeterminism.

```
type Nondet' a = (Fail :+: Alt) a

list :: Free Nondet' a → [a]

list = eval alg gen where
   gen :: a → [a]
   gen x = [x]

alg :: Nondet' [a] → [a]
   alg (L Fail) = []
   alg (R (Alt k)) = k True ++ k False

-- k :: Bool → [a]
```

This demonstrates that the parameter to a syntax functor sometimes has the form of a function, i.e. Bool  $\rightarrow$  k.

#### 3.6 State

A stateful computation can be modelled by having two operations, Get and Put.

```
data State s k = Put s k
| Get (s \rightarrow k)
```

The intuition is that Put s k will put the value s into the state before continuing with the computation k. The Get f operation will only continue when  $f :: s \to k$  is given a variable of type s. The semantic domain for State is a function of type:

```
s \rightarrow (a,s)
```

This is a carrier for stateful computations.

```
evalState :: Free (State s) a \rightarrow (s \rightarrow (a,s))

evalState = eval alg gen where

gen :: a \rightarrow (s \rightarrow (a,s))

gen x s = (a,s)

-- or equivalently gen x = \lambdas \rightarrow (x,s)

alg :: State s (s \rightarrow (a,s)) \rightarrow (s \rightarrow (a,s))

alg (Put s' k) = \lambdas \rightarrow k s'

alg (Get k) = \lambdas \rightarrow k s s

-- in k s s, the first s is the state that generates programs, and the second is the state parsed on to future programs.
```

This function carries on computations generated by k when supplied with the new state s.

### 3.7 Diagrams of operations

We have already seen many diagrams, here are some conventions:



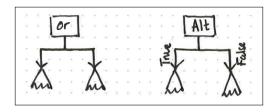
- ullet Operations are written in boxes, and the arrows emerging below represent different k values.
- Trianglular leaves represent variables.



• We will often represent an arbitrary subtree with this.

To represent nondeterminism and alternation, we have these diagrams:

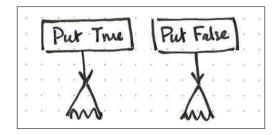
```
data Or k = Or k k
data Alt k = Alt (Bool \rightarrow k)
```



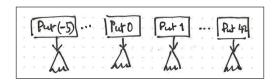
We can imagine that State s = Put s :+: Get s:

```
data Put s k = Put s k data get s k = Get (s \rightarrow k)
```

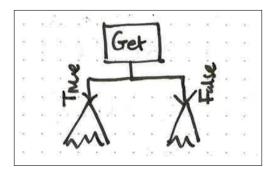
To draw the operation Put, we have the following:



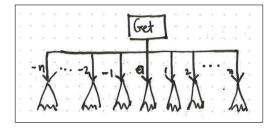
These examples are of Free (Put Bool) a syntax trees: Since s = Bool we can only construct these two Put nodes. We cab parameterise s different, so if we had s = Int, we would have a huge (infinite) number of nodes:



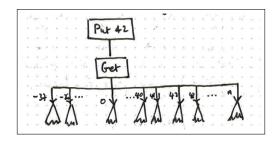
The Get nodes are a generalisation of Alt. If we consider trees of the form Free (Get Bool) athen this is as follows:



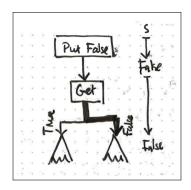
However, with s = Int, we have Free (Get Int) a:



Note that Alt and Get Bool are not syntactically very similar; they share the same structure but their differences are exhibited when we provide different algebras. Another thing we can do is compose trees together:



When we gave the algebra for State s when the carrier was  $s \rightarrow (a,s)$ , we were storing the state s in output, and reacting as s as input. Consider s = Bool:



Note that here we have used False to choose which child to use, and we also passed False on to the next stage.

```
alg (Get k) = \lambdas \rightarrow k s s
```

- -- The k in 'Get k' is represented by the children in the diagram.
- -- The first s in 'k s s' is selecting the child.
- -- The second s is passing the s onwards.

These syntax trees all correspond to values of type Free (State s) a. They can be cumbersome to work with because we must wrap everything in Op.

```
Op (Put False (Op (Get (\lambda s \rightarrow ...))))
```

We can avoid this by introducing smart constructors for Put and Get:

```
put :: s \rightarrow Free (State s) ()

put s = op (Put s (Var ()))

get :: Free (State s) s

get = Op (Get (\lambda s \rightarrow Var s))
```

And with these tools, we can simply substitute

```
put False » get
```

and this produces the same tree as above.