# Folding Unfolded Polyglot FP for Fun and Profit Haskell and Scala

See how recursive functions and structural induction relate to recursive datatypes

Follow along as the **fold abstraction** is introduced and explained

Watch as folding is used to simplify the definition of recursive functions over recursive datatypes

Part 1 - through the work of



Richard Bird http://www.cs.ox.ac.uk/people/richard.bird/



Graham Hutton

A tutorial on the universality and expressiveness of fold

> GRAHAM HUTTON University of Nottingham, Nottingham, UK http://www.cs.nott.ac.uk/-gmh







This slide deck is almost entirely centered on material from **Richard Bird**'s fantastic book, **Introduction to Functional Programming using Haskell**.

I hope he'll forgive me for relying so heavily on his work, but I couldn't resist using extensive excerpts from his compelling book to introduce and explain the concept of folding.





Richard Bird http://www.cs.ox.ac.uk/people/richard.bird/

https://en.wikipedia.org/wiki/Richard\_Bird\_(computer\_scientist)

#### **3.1 Natural Numbers**

The natural numbers are the numbers 0, 1, 2 and so on, used for counting. The type *Nat* is introduced by the declaration

data Nat = Zero | Succ Nat

*Nat* is our first example of a **recursive datatype declaration**. The definition says that *Zero* is a value of *Nat*, and that *Succ n* is a value of *Nat* whenever *n* is. In particular, the constructor *Succ* (short for 'successor'), has type *Nat*  $\rightarrow$  *Nat*. For example, each of

Zero, Succ Zero, Succ (Succ Zero)

is an element of *Nat*. As an element of *Nat* the number 7 would be represented by

```
Succ (Succ (Succ (Succ (Succ (Succ Zero))))))
```

Every natural number is represented by a unique value of *Nat*. On the other hand, not every value of *Nat* represents a well-defined natural number. In fact *Nat* also contains the values  $\perp$ , *Succ*  $\perp$ , *Succ* (*Succ*  $\perp$ ), and so on. These additional values will be discussed later.

Let us see how to program the basic arithmetic and comparison operations on *Nat*. Addition can be defined by

(+) ::  $Nat \rightarrow Nat \rightarrow Nat$  m + Zero = mm + Succ n = Succ (m + n)

This is a **recursive definition**, defining + by pattern matching on the second argument. Since every element of *Nat*, apart for  $\bot$ , is either *Zero* or of the form *Succ n*, where *n* is an element of *Nat*, the two patterns in the equations for + are disjoint and cover all numbers apart from  $\bot$ .





**Richard Bird** 

•••

Here is how **Zero + Succ** (Succ Zero) would be evaluated:

Zero + Succ (Succ Zero)

- = { second equation for +, i.e. m + Succ n = Succ (m + n) } Succ (Zero + Succ Zero)
- = { second equation for +, i.e. m + Succ n = Succ (m + n) } Succ (Succ (Zero + Zero))
- = { first equation for +, i.e. m + Zero = m }
  Succ (Succ Zero)

...it is not a practical proposition to introduce natural numbers through the datatype Nat: arithmetic would be just too inefficient. In particular, calculating m + n would require (n + 1) evaluation steps. On the other hand, counting on your fingers is a good way to understand addition.

Given +, we can define ×:

Given  $\times$ , we can define exponentiation (1) by

```
(†) :: Nat \rightarrow Nat \rightarrow Nat

m \uparrow Zero = Succ Zero

m \uparrow Succ n = (m \uparrow n) \times m
```





•••



On the next slide we show the definitions of +,  $\times$ , and  $\uparrow$  again, and have a go at implementing the three operations in Scala, together with some tests.

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

#### sealed trait Nat

case class Succ(n: Nat) extends Nat
case object Zero extends Nat

$(+) \qquad ::  Nat \to Nat \to Nat$ $m + Zero = m$ $m + Succ n = Succ (m + n)$	$(\times) \qquad ::  Nat \to Nat \to Nat$ $m \times Zero \qquad = \qquad Zero$ $m \times Succ n \qquad = \qquad (m \times n) + m$	$(\uparrow) \qquad ::  Nat \to Nat \to Nat$ $m \uparrow Zero \qquad =  Succ Zero$ $m \uparrow Succ n \qquad =  (m \uparrow n) \times m$	
<pre>val `(+)`: Nat =&gt; Nat =&gt; Nat =     m =&gt; {         case Zero =&gt; m         case Succ(n) =&gt; Succ(m + n)         }</pre>	<pre>val `(x)`: Nat =&gt; Nat =&gt; Nat =     m =&gt; {         case Zero =&gt; Zero         case Succ(n) =&gt; (m × n) + m     }</pre>	<pre>val `(^)`: Nat =&gt; Nat =&gt; Nat =     m =&gt; {         case Zero =&gt; Succ(Zero)         case Succ(n) =&gt; (m ^ n) × m     }</pre>	<pre>implicit class NatOps(m: Nat){     def +(n: Nat) = `(+)`(m)(n)     def ×(n: Nat) = `(×)`(m)(n)     def ↑(n: Nat) = `(↑)`(m)(n) }</pre>

; assert(Math.pow(1,0) == 1) ; assert( (Succ(Zero) ↑ Zero) == Succ(Zero) )
; assert(Math.pow(2,2) == 4) ; assert( (Succ(Succ(Zero)) ↑ Succ(Succ(Zero))) == Succ(Succ(Succ(Succ(Zero)))) )

The remaining arithmetic operation common to all numbers is subtraction (-). However, subtraction is a **partial operation** on natural numbers. The definition is

This definition uses pattern matching on both arguments; taken together, the patterns are disjoint but not exhaustive. For example,

```
Succ Zero - Succ (Succ Zero)
= { second equation for -, i.e. Succ m - Succ n = m - n }
Zero - Succ Zero
= { case exhaustion }
⊥
```

The hint 'case exhaustion' in the last step indicates that no equation for (-) has a pattern that matches (*Zero* – *Succ Zero*). More generally,  $m - n = \bot$  if m < n. The **partial nature** of subtraction on the natural numbers is the prime motivation for introducing the **integer** numbers; over the **integers**, (-) is a **total operation**. *...*Finally, here are two more examples of programming with *Nat*. The **factorial** and **Fibonacci** functions are defined by

```
fact :: Nat \rightarrow Nat
fact Zero = Succ Zero
fact (Succ n) = Succ n \times fact n
fib :: Nat \rightarrow Nat
fib Zero = Zero
fib (Succ Zero) = Succ Zero
fib (Succ (Succ n)) = fib (Succ n) + fib n
```







See the next slide for a Scala implementation of the – operation.





On the next slide we show the definitions of fact, and fib again, and have a go at implementing the two functions in Scala, together with some tests.

fact	::	$Nat \rightarrow Nat$
fact <mark>Zero</mark>	=	Succ Zero
fact (Succ n)	=	Succ $n \times fact n$

```
val fact: Nat => Nat = {
  case Zero => Succ(Zero)
  case Succ(n) => Succ(n) × fact(n)
}
```

fib	::	$Nat \rightarrow Nat$
fib <mark>Zero</mark>	=	Zero
fib ( <mark>Succ Zero</mark> )	=	Succ Zero
$fib\left(Succ\left(Succn\right)\right)$	=	fib(Succn) + fibn

<pre>val fib: Nat =&gt; Nat =</pre>	- {
case Zero	=> Zero
case Succ(Zero)	=> Succ(Zero)
<pre>case Succ(Succ(n))</pre>	=> fib(Succ(n)) + fib(n)
1	

#### 3.1.1 Partial numbers

Let us now return to the point about there being extra values in *Nat*. The values

```
\perp, Succ \perp, Succ (Succ \perp), ...
```

are all different and each is also a member of *Nat*. That they exist is a consequence of three facts:

- *i.*  $\perp$  is an element of *Nat* because every datatype declaration introduces at least one extra value, the **undefined** value of the type.
- ii. constructor functions of a datatype are assumed to be **nonstrict**
- *iii.* **Succ** *n* is an element of **Nat**, whenever *n* is

To appreciate why these extra values are different from one another, suppose we define undefined :: Nat by the equation undefined = undefined. Then

```
? Zero < undefined
{Interrupted!}
? Zero < Succ undefined
True
? Succ Zero < Succ undefined
{Interrupted!}
? Succ Zero < Succ (Succ undefined)
True</pre>
```

One can interpret the extra values in the following way:  $\bot$  corresponds to the natural number about which there is absolutely no information; *Succ*  $\bot$  to the natural number about which the only information is that it is greater than *Zero*; *Succ* (*Succ*  $\bot$ ) to the natural number about which the only information is that it is greater than *Succ Zero*; and so on.





There is also one further value of *Nat*, namely the 'infinite' number:

```
Succ (Succ (Succ (Succ ...)))
```

This number can be defined by

infinity :: Nat infinity = Succ infinity

It is different from all the other numbers, because it is the only number *x* for which *Succ m* < *x* returns *True* for all finite numbers *m*. In this sense, *inf inity* is the largest element of *Nat*. If we request the value of *inf inity*, then we obtain

```
? infinity
Succ (Succ (Succ (Succ [Interrupted!]))
```

The number *infinity* satisfies other properties, in particular n + infinity = infinity, for all numbers n. The dual equation *infinity* + n = infinity holds only for finite numbers n. We will see how to prove assertions such as these in the next section.

To summarise this discussion, we can divide the values of *Nat* into three classes:

- The **finite** numbers, those that correspond to well-defined natural numbers.
- The **partial** numbers,  $\perp$ , **Succ**  $\perp$ , and so on.

...

• The **infinite** numbers, of which there is just one, namely *infinity*.

We will see that this classification holds true of *all* recursive types. There will be the finite elements of the type, the partial elements, and the infinite elements. <u>Although the infinite natural number is not of much use, the same is not true of the infinite values of other datatypes</u>.







Note that when in this slide deck we mention the concepts of  $\bot$  and *infinity*, it is mainly in a **Haskell** context, as we did in the last two slides. In particular, we won't be modelling  $\bot$  and *infinity* in any of the Scala code you'll see throughout the deck.

# **3.2 Induction**

In order to reason about the properties of **recursively defined** functions over a **recursive datatype**, we can appeal to a principle of **structural induction**. In the case of **Nat**, the principle of **structural induction** can be defined as follows: In order to show that some property P(n) holds for each finite number n of **Nat**, it is sufficient to show:

**Case** (*Zero*). That P(Zero) holds.

**Case** (*Succ* n). That if P(n) holds, then P(Succ n) holds also.

Induction is valid for the same reason that recursive definitions are valid: every finite number is either *Zero* or of the form *Succ n*, where *n* is a finite number. If we prove the first case, then we have shown that the property is true for *Zero*; If we also prove the second case, then we have shown that the property is true for *Succ Zero*, since it is true for *Zero*. But now, by the same argument, it is true for *Succ (Succ Zero)*, and so on.

The principle needs to be extended if we want to assert that some proposition is true for *all* elements of *Nat*, but we postpone discussion of this point for the following section.

As an example, let's prove that Zero + n = n for all finite numbers *n*. Recall that + is defined by

m + Zero = mm + Succ n = Succ (m + n)

The first equation asserts that **Zero** is a right unit of +. In general, *e* is a **left unit** of  $\oplus$  if  $e \oplus x = x$  for all *x*, and a **right unit** of *x* if  $x \oplus e = x$  for all *x*. If *e* is both a **left unit** and a **right unit** of an operator  $\oplus$ , then it is called *the* **unit** of  $\oplus$ . The terminology is appropriate since only one value can be both a left and right **unit**. So, by proving that **Zero** is a **left unit**, we have proved that **Zero** is *the* **unit** of +.





**Proof**. The proof is by induction on *n*. More precisely, we take for P(n) the assertion that Zero + n = n. This equation is referred to as the **induction hypothesis**.

**Case** (*Zero*). We have to show Zero + Zero = Zero, which is immediate from the first equation defining +.

**Case** (*Succ* n). We have to show that Zero + Succ n = Succ n, which we do by simplifying the left-hand expression:

Zero + Succ n

- = { second equation for +, i.e. m + Succ n = Succ (m + n) } Succ (Zero + n)
- = { induction hypothesis}
   Succ n

This example shows the format we will use for **inductive proofs**, laying out each case separately and using a  $\Box$  to mark the end. The very last step made use of the **induction hypothesis**, which is allowed by the way **induction** works.

# 3.2.1 Full Induction

In the form given above, the **induction principle** for *Nat* suffices only to prove properties of the **finite** members of *Nat*. If we want to show that a property *P* also hold for every **partial number**, then we have to prove three things:

**Case** ( $\perp$ ). That  $P(\perp)$  holds.

**Case** (*Zero*). That P(Zero) holds.

**Case** (*Succ* n). That if P(n) holds, then P(Succ n) holds also.

We can omit the second case, but then we can conclude only that P(n) holds for every **partial** number. The reason the principle is valid is that is that every **partial number** is either  $\perp$  or of the form *Succ n* for some **partial number** *n*.





To illustrate, let us prove the somewhat counterintuitive result that m + n = n for all numbers m and all partial numbers n.

**Proof**. The proof is by **partial number induction** on *n*.

**Case** ( $\perp$ ). The equation  $m + \perp = \perp$  follows at once by case exhaustion in the definition of +. That is,  $\perp$  does not match either of the patterns *Zero* or *Succ n*.

Case (Succ n). For the left-hand side, we reason

```
m + Succ n
```

- = { second equation for +, i.e. m + Succ n = Succ (m + n) } Succ (m + n)
- = { induction hypothesis} *Succ* n

Since the right-hand side is also *Succ n*, we are done.

# 3.2.2 Program synthesis

In the proofs above we defined some functions and then used induction to prove a certain property. We can also view induction as a way to **synthesise** definitions of functions so that they satisfy the properties we want.

Let us illustrate with a simple example. Suppose we **specify** subtraction of natural numbers by the condition

(m+n) - n = m

for all m and n. The specification does not give a constructive definition of (-), merely a property that it has to satisfy. However, we can do an **induction proof** on n of the equation above, but view the calculation as a way of generating a suitable definition of (-).





Unlike previous proofs, we reason with the equation as a whole, since simplification of both sides independently is not possible if we do not know what all the rules of simplification are.

Case (Zero). We reason

(m + Zero) - Zero = m  $\equiv \{ \text{first equation for +, i.e. } m + Zero = m \} \}$ m - Zero = m

Hence we can take m - Zero = m to satisfy the case. The symbol  $\equiv$  is used to separate steps of the calculation since we are calculating with mathematical assertions, not with values of a datatype.

Case (Succ n). We reason

```
(m + Succ n) - Succ n = m
\equiv \{ \text{second equation for +, i.e. } m + Succ n = Succ (m + n) \} \}
Succ (m + n) - Succ n = m
\equiv \{ \text{hypothesis } (m + n) - n = m \}
Succ (m + n) - Succ n = (m + n) - n
```

Replacing m + n in the last equation by m, we can take Succ m - Succ n = m - n to satisfy the case. Hence we have derived

```
m - Zero = m
Succ m - Succ n = m - n
```

This is the program for (-) seen earlier.







After that look at **structural induction**, it is finally time to see how **Richard Bird** introduces the concept of **folding**.

# 3.3 The fold function

Many of the **recursive definitions** seen so far have a common pattern, exemplified by the following definition of a function *f* :

 $f :: Nat \to A$ f Zero = cf (Succ n) = h (f n)

Here, *A* is some type, *c* is an element of *A*, and  $h :: A \to A$ . Observe that *f* works by taking an element of *Nat* and replacing *Zero* by *c* and *Succ* by *h*. For example, *f* takes

```
Succ (Succ (Succ Zero)) to h(h(hc))
```

The two equations for *f* can be captured in terms of a single function, *foldn*, called the *fold* function for *Nat*. The definition is

```
 \begin{array}{ll} foldn & :: & (\alpha \rightarrow \alpha) \rightarrow \alpha \rightarrow \textit{Nat} \rightarrow \alpha \\ foldn \ h \ c \ \textit{Zero} & = \ c \\ foldn \ h \ c \ \textit{(Succ } n) & = \ h \ (foldn \ h \ c \ n) \end{array}
```

In particular, we have

m + n = foldn Succ m n  $m \times n = foldn (+ m) Zero n$  $m \uparrow n = foldn (\times m) (Succ Zero) n$ 

It follows also that the **identity function** *id* on *Nat* satisfies id = foldn Succ Zero. A suitable <u>fold</u> function can be defined for <u>every recursive type</u>, and we will see other <u>fold</u> functions in the following chapters.





**Richard Bird** 



Just to reinforce the ideas on the previous slide, here are the original definitions of +,  $\times$  and  $\uparrow$ , and next to them, the new definitions in terms of *foldn*.

And the next slide is the same but in terms of Scala code.



$$\begin{array}{ll} foldn & :: & (\alpha \rightarrow \alpha) \rightarrow \alpha \rightarrow \textit{Nat} \rightarrow \alpha \\ foldn \ h \ c \ \textit{Zero} & = & c \\ foldn \ h \ c \ (\textit{Succ} \ n) & = & h \ (foldn \ h \ c \ n) \end{array}$$



```
def foldn[A](h: A => A, c: A, n: Nat): A =
    n match {
    case Zero => c
    case Succ(n) => h(foldn(h,c,n))
  }
```

In the examples above, each instance of *foldn* also returned an element of *Nat*. In the following two examples, *foldn* returns an element of (*Nat*, *Nat*):

```
fact :: Nat \rightarrow Nat

fact = snd \cdot foldn f (Zero, Succ Zero)

where f(m, n) = (Succ m, Succ (m) \times n)
```

```
\begin{array}{rcl} fib & :: & \textit{Nat} \rightarrow \textit{Nat} \\ fib & = & fst \cdot \textit{foldn} \ g \ (\textit{Zero}, \textit{Succ Zero}) \\ & & \text{where} \ g(m,n) = (n,m+n) \end{array}
```

The function *fact* computes the **factorial** function and function *fib* computes the **Fibonacci** function. Each program works by first computing a more general result, namely an element of (*Nat*, *Nat*), and then extracts the required result. In fact,

foldn f (Zero, Succ Zero) n = (n, fact n)foldn g (Zero, Succ Zero) n = (fib n, fib (Succ n))

These equations can be **proved by induction**. The program for fib is more efficient than a direct **recursive definition**. The recursive program requires an **exponential number** of + operations, while the program above requires only a **linear number**. We will discuss efficiency in more detail in chapter 7, where the programming technique that led to the invention of the new program for fib will be studied in a more general setting.

There are two advantages of writing **recursive definitions** in terms of *foldn*. Firstly, the definition is shorter; rather than having to write down two equations, we have only to write down one. Secondly, it is possible to prove general properties of *foldn* and use them to prove properties of specific instantiations. In other words, rather than having to write down many **induction proofs**, we have only to write down one.







The next slide shows the original definitions of the **factorial** and **Fibonacci** functions, and next to them, the new definitions in terms of *foldn*.

And the slide after that is the same but in terms of Scala code.

fact	::	$Nat \rightarrow Nat$
fact <mark>Zero</mark>	=	Succ Zero
fact ( <mark>Succ</mark> n)	=	<mark>Succ</mark> n × fact n



ıct	::	$Nat \rightarrow Nat$
ıct	=	snd • foldn f (Zero, Succ Zero)
		where $f(m, n) = (Succ m, Succ (m) \times n)$



foldn	::	$(\alpha \rightarrow \alpha) \rightarrow \alpha \rightarrow Nat \rightarrow \alpha$
foldn h c <mark>Zero</mark>	=	С
foldn h c (Succ n)	=	h ( <mark>foldn</mark> h c n)

```
val fact: Nat => Nat = {
   case Zero => Succ(Zero)
   case Succ(n) => Succ(n) × fact(n)
}
```



```
def fact(n: Nat): Nat = {
    def snd(pair: (Nat, Nat)): Nat =
    pair match { case (_,n) => n }
    def f(pair: (Nat, Nat)): (Nat, Nat) =
    pair match { case (m,n) => (Succ(m), Succ(m) × n) }
```

```
snd( foldn(f, (Zero, Succ(Zero)), n) )
```

<pre>val fib: Nat =&gt; Nat =</pre>	= {	
case Zero	=>	Zero
<pre>case Succ(Zero)</pre>	=>	Succ(Zero)
<pre>case Succ(Succ(n))</pre>	=>	<pre>fib(Succ(n)) + fib(n)</pre>
}		



```
def fib(n: Nat): Nat = {
    def fst(pair: (Nat, Nat)): Nat =
        pair match { case (n,_) => n }
    def g(pair: (Nat, Nat)): (Nat, Nat) =
        pair match { case (m,n) => (n, m + n) }
    fst( foldn(g, (Zero, Succ(Zero)), n) )
```

```
def foldn[A](h: A => A, c: A, n: Nat): A =
    n match {
    case Zero => c
    case Succ(n) => h(foldn(h,c,n))
  }
```



Now let's have a very quick look at the datatype for lists, and at induction over lists.

#### 4.1.1 Lists as a datatype

A **list** can be constructed from scratch by starting with the **empty list** and successively adding elements one by one. One can add elements to the front of the list, or to the rear, or to somewhere in the middle. In the following datatype declaration, nonempty lists are constructed by adding elements to the front of the list:

```
data List \alpha = Nil | Cons \alpha (List \alpha)
```

...The constructor (short for 'construct' – the name goes back to the programming language LISP) adds an element to the front of the list. For example, the list [1,2,3] would be represented as the following element of *List Int*:

```
Cons 1 (Cons 2 (Cons 3 Nil))
```

In functional programming, lists are defined as elements of *List*  $\alpha$ . The syntax [ $\alpha$ ] is used instead of *List*  $\alpha$ , the constructor *Nil* is written as [], and the constructor *Cons* is written as infix operator (:). Moreover, (:) associates to the right, so

[1,2,3] = 1:(2:(3:[])) = 1:2:3:[]

In other words, the special syntax on the left can be regarded as an abbreviation for the syntax on the right, which is also special, but only by virtue of the fact that the constructors are given nonstandard names.

Like functions over other datatypes, functions over lists can be defined by pattern matching.







Before moving on to the topic of **induction over lists**, **Richard Bird** gives an example of a function defined over **lists** using **pattern matching**, but the function he chooses is the **equality function**, whereas we are going to choose the **sum function**, just to keep things simpler.

sum	::	[Int] → Int
sum []	=	0
sum(x:xs)	=	x + (sum xs)

assert( sum( 1 :: (2 :: (3 :: Nil)) ) == 6)

6)



Same as on the previous slide, but this time using **Nat** rather than **Int**, just for fun.

data List  $\alpha$  = Nil | Cons  $\alpha$  (List  $\alpha$ )

data Nat = Zero | Succ Nat

(+)	::	$Nat \rightarrow Nat \rightarrow Nat$
m <b>+ Zero</b>	=	m
т <b>+ Succ</b> п	=	<b>Succ</b> $(m+n)$

sum	::	List Nat $\rightarrow$ Nat
sum <mark>Nil</mark>	=	Zero
sum Cons x xs	=	x + (sum xs)

```
sealed trait Nat
case class Succ(n: Nat) extends Nat
case object Zero extends Nat
```

```
sealed trait List[+A]
case class Cons[+A](head: A, tail: List[A]) extends List[A]
case object Nil extends List[Nothing]
```

```
val `(+)`: Nat => Nat => Nat =
    m => { case Zero => m
        case Succ(n) => Succ(m + n) }
```

```
implicit class NatSyntax(m: Nat){
   def +(n: Nat) = `(+)`(m)(n)
```

```
val sum: List[Nat] => Nat = {
    case Nil => Zero
    case Cons(x, xs) => x + sum(xs)
```

#### 4.1.2 Induction over Lists

Recall from section **3.2** that, for the datatype *Nat* of natural numbers, structural induction is based on three cases: every element of *Nat* is either  $\bot$ , or *Zero*, or else has the form *Succ n* for some element *n* of *Nat*. Similarly, structural induction on lists is also based on on three cases: every list is either the undefined list  $\bot$ , the empty list [], or else has the form (*x*: *xs*) for some *x* and list *xs*.

To show by induction that a proposition P(xs) holds for all lists xs it suffices therefore to establish three cases:

**Case** ( $\perp$ ). That  $P(\perp)$  holds.

**Case** ([]). That *P*([]) holds.

**Case** (x: xs). That if P(xs) holds, then P(x: xs) also holds for every x.

If we prove only the second two cases, then we can conclude only that P(xs) holds for every **finite** list; if we prove only the first and third cases. Then we can conclude only that P(xs) holds for every **partial** list. If P takes the form of an equation, as all of our laws do, then proving the first and third cases is sufficient to show that P(xs) holds for every **infinite** list. **Partial** lists and **infinite** lists are described in the following section. Examples of induction proofs are given throughout the remainder of the chapter.







**Richard Bird** provides other examples of **recursive functions** over **lists**. Let's see some of them: list concatenation, flattening of lists of lists, list reversal and length of a list.

When looking at the first one, i.e. concatenation, let's also see an example of **proof by structural induction** on **lists**.

### 4.2.1 Concatenation

Two lists can be **concatenated** to form one longer list. This function is denoted by the binary operator **#** (pronounced 'concatenate'). As two simple examples, we have

?[1,2,3] **#** [4,5] [1,2,3,4,5]

?[1,2] **#** [] **#** [1] [1,2,1]

The formal definition of **#** is

 $(\texttt{#}) \qquad :: \quad [\alpha] \to [\alpha] \to [\alpha]$  $[] \texttt{#} ys \qquad = \quad ys$  $(x:xs) \texttt{#} ys \qquad = \quad x: (xs \texttt{#} ys)$ 

**Concatenation** takes two lists, both of the same type, and produces a third list, again of the same type. Hence the type assignment. The definition of # is by pattern matching on the left-hand argument; the two patterns are disjoint and cover all cases, apart from the undefined list  $\bot$ . It follows by case exhaustion that  $\bot \# ys = \bot$ .

However, it is not the case that  $y_s \# \bot = \bot$ . For example,

?[1,2,3] **#** *undefined* [1,2,3{*Interrupted*!}]

The list  $[1,2,3] \# \perp$  is a **partial** list; In full form it is the list  $1:2:3:\perp$ . The evaluator can compute the first three elements, but thereafter it goes into a **nonterminating** computation, so we interrupt it.

The second equation for *H* is very succinct and requires some thought. Once one has come to grips with the definition of *H*, one





has understood a good deal about how lists work in functional programming. Note that the number of steps required to compute xs + ys is proportional to the number of elements in xs.

[1, 2] **#** [3, 4, 5] { notation} = (1:(2:[]) + (3:(4:(5:[]))){ second equation for #, i.e. (x:xs) # ys = x : (xs # ys) } = 1:((2:[]) + (3:(4:(5:[])))){ second equation for # } = 1:(2:([] + (3:(4:(5:[])))))){ first equation for # i.e., [] # ys = ys } = 1:(2:(3:(4:(5:[])))){ notation} = [1, 2, 3, 4, 5]

Concatenation is an associative operation with unit []:

(xs + ys) + zs = xs + (ys + zs)xs + [] = [] + xs = xs

Let us now prove by induction that **#** is **associative**.

**Proof**. The proof is by **induction** on *xs*. **Case** ( $\perp$ ). For the left-hand side, we reason

 $\bot \# (ys \# zs)$ 

- = { case exhaustion}  $\perp \# zs$
- = { case exhaustion}





The right-hand side simplifies to  $\bot$  as well, establishing the case.

Case ([]). For the left hand side, we reason

```
[] # (ys # zs)
= { first equation for # i.e. , [] # ys = ys }
(ys # zs)
```

The right-hand side simplifies to (ys + zs) as well, establishing the case.

```
Case (x : xs). For the left hand side, we reason
```

```
((x : xs) # ys) # zs
```

- = { second equation for #, i.e. (x:xs) # ys = x : (xs # ys) } (x : (xs # ys)) # zs
- = { second equation for **#** }
  - x:((xs + ys) + zs)
- = { induction hypothesis }
  - x:(xs + (ys + zs))

For the right-hand side we reason

(x:xs) # (ys # zs) $= \{ second equation for #, i.e. (x : xs) # ys = x : (xs # ys) \}$ x : (xs # (ys # zs))

The two sides are equal, establishing the case.

...Note that associativity is proved for *all* lists, **finite**, **partial** or **infinite**. Hence we can assert that **#** is **associative** without qualification....





### 4.2.2 Concat

**Concatenation** performs much the same function for lists as the **union operator** U does for sets. A companion function is *concat*, which concatenates a list of lists into one long list. This function, which roughly corresponds to the **big-union operator** U for sets of sets, is defined by

```
concat::[[\alpha]] \rightarrow [\alpha]concat[]=concat(xs:xss)=xs + concat xss
```

For example,

```
?concat [[1,2],[],[3,2,1]]
[1,2,3,2,1]
```

### 4.2.3 Reverse

Another basic function on lists is *reverse*, the function that reverses the order of elements in a finite list. For example:

```
? reverse "Madam, I'm Adam."
".MadA m'I ,madaM"
```

The definition is

```
reverse::[\alpha] \rightarrow [\alpha]reverse[]=reverse(x:xs)=reverse(x:xs)=
```

In words, to reverse a list (x : xs), one reverses xs, and then adds x to the end. As a program, the above definition is not very





efficient: on a list of length n, it will need a number of reduction steps proportional to  $n^2$  to deliver the reversed list. The first element will be appended to the end of a list of length (n - 1), which will take about (n - 1) steps, the second element will be appended to a list of length (n - 2), taking (n - 2) steps, and so on. The total time is therefore about

 $(n - 1) + (n - 2) + \dots 1 = n(n - 1)/2$  steps

A more precise analysis is given in chapter 7, and a more efficient program for *reverse* is given in section 4.5.

## 4.2.2 Length

The length of a list is the number of elements it contains:

 $\begin{array}{rcl} length & :: & [\alpha] \rightarrow Int \\ length [] & = & 0 \\ length (x:xs) & = & 1 + length xs \end{array}$ 

The nature of the list element is irrelevant when computing the length of a list, whence the type assignment. For example,

```
? length [undefined, undefined]
```

2

However, not every list has a well-defined length. In particular, the **partial** lists  $\bot$ ,  $x : \bot$ ,  $x : y : \bot$ , and so on, have an undefined length. Only **finite** lists have well-defined lengths. The list  $[\bot, \bot]$  is a **finite** list, *not* a **partial** list, because it is the list  $\bot : \bot : []$ , which ends in [], not  $\bot$ . The computer cannot produce the elements, but it can produce the length of the list.





# 4.3 Map and filter

Two useful functions on lists are *map* and *filter*. The function *map* applies a function to each element of a list. For example

? *map square* [9, 3] [81, 9] ? *map nextLetter* "HAL" "IBM"

The definition is

$$\begin{array}{rcl} map & :: & (\alpha \to \beta) \to [\alpha] \to [\beta] \\ map \ f \ [ \ ] & = & [ \ ] \\ map \ f \ (x : xs) & = & f \ x : map \ f \ xs \\ ... \end{array}$$

# 4.3 filter

•••

The second function, *filter*, takes a Boolean function *p* and a list *xs* and returns that sublist of *xs* whose elements satisfy *p*. For example,

 ? filter even [1,2,4,5,32]
 ? (sum · map square · filter even) [1..10]

 [2,4,32]
 220

The last example asks for the sum of the squares of the even integers in the range 1..10.

The definition of filter is

```
filter::(\alpha \rightarrow Bool) \rightarrow [\alpha] \rightarrow [\alpha]filter p []=[]filter p (x : xs)=if p x then x : filter p xs else filter p xs
```







Now let's look at **fold** functions over lists.

# 4.5 The fold functions

We have seen in the case of the datatype *Nat* that many **recursive definitions** can be expressed very succinctly using a suitable *fold* operator. Exactly the same is true of lists. Consider the following definition of a function *h* :

 $\begin{array}{l} h \left[ \right] &= e \\ h \left( x : xs \right) &= x \bigoplus h \, xs \end{array}$ 

The function h works by taking a list, replacing [] by e and (:) by  $\bigoplus$ , and evaluating the result. For example, h converts the list

 $x_1:(x_2:(x_3:(x_4:[])))$ 

to the value

 $x_1 \oplus (x_2 \oplus (x_3 \oplus (x_4 \oplus e)))$ 

Since (:) associates to the right, there is no need to put in parentheses in the first expression. However, we do need to put in parentheses in the second expression because we do not assume that  $\bigoplus$  associates to the right.

The pattern of definition given by *h* is captured in a function *foldr* (prounced 'fold right') defined as follows:

 $\begin{array}{ll} foldr & :: (\alpha \rightarrow \beta \rightarrow \beta) \rightarrow \beta \rightarrow [\alpha] \rightarrow \beta \\ foldr \ f \ e \ [ \ ] & = e \\ foldr \ f \ e \ (x: xs) & = f \ x \ (foldr \ f \ e \ xs) \end{array}$ 

We can now write h = foldr ( $\bigoplus$ ) *e*. The first argument of *foldr* is a binary operator that takes an  $\alpha$ -value on its left and an a  $\beta$ -value on its right, and delivers a  $\beta$ -value. The second argument of *foldr* is a  $\beta$ -value. The third argument is of type [ $\alpha$ ], and the result is of type  $\beta$ . In many cases,  $\alpha$  and  $\beta$  will be instantiated to the same type, for instance when  $\bigoplus$  denotes an **associative operation**.







In the next slide we look at how some of the recursively defined functions on lists that we have recently seen can be redefined in terms of foldr.

To aid comprehension, I have added the original function definitions next to the new definitions in terms of *foldr*. For reference, I also added the definition of *foldr*.

The single function *foldr* can be used to define almost every function on lists that we have met so far. Here are just some examples:

 $:: [[\alpha]] \rightarrow [\alpha]$ concat = foldr(#)[]concat  $:: [\alpha] \rightarrow [\alpha]$ reverse reverse = foldr snoc [] where snoc x xs = xs + [x]*length*  $:: [\alpha] \rightarrow Int$ length = foldr oneplus 0 where oneplus x n = 1 + n...  $:: [Int] \to Int$ sum = foldr (+) 0 sum  $:: (\alpha \to \beta) \to [\alpha] \to [\beta]$ map  $= foldr (cons \cdot f) []$ map f where cons x xs = x : xs...

<i>concat</i> <i>concat</i> [] <i>concat</i> (xs:xss)	$ \begin{array}{l} ::  [[\alpha]] \to [\alpha] \\ =  [] \\ =  xs \ \# \ concat \ xss \end{array} $
reverse reverse[] reverse(x : xs)	:: $[\alpha] \rightarrow [\alpha]$ = [] = reverse xs + [x]
length length[] length (x:xs)	$\begin{array}{ll} :: & [\alpha] \rightarrow Int \\ = & 0 \\ = & 1 + length  xs \end{array}$
sum sum[] sum (x: xs)	$\begin{array}{ll} :: & [Int] \rightarrow Int \\ = & 0 \\ = & x + (sum \ xs) \end{array}$
<i>map</i> <i>map f</i> [] <i>map f (x : xs)</i>	$\begin{array}{ll} :: & (\alpha \to \beta) \to [\alpha] \to [\beta] \\ = & [] \\ = & f \ x : map \ f \ xs \end{array}$

foldr::  $(\alpha \rightarrow \beta \rightarrow \beta) \rightarrow \beta \rightarrow [\alpha] \rightarrow \beta$ foldr f e []= efoldr f e (x:xs)= f x (foldr f e xs)







**C** On the next slide, the same code translated into **Scala** 

**@philip\_schwarz** 

```
:: (\alpha \to \beta \to \beta) \to \beta \to [\alpha] \to \beta
foldr
foldr f e []
                      = e
foldr f e(x:xs) = f x (foldr f e xs)
(#) :: [\alpha] \rightarrow [\alpha] \rightarrow [\alpha]
\left[\right] + ys
           = ys
(x:xs) + ys = x:(xs + ys)
concat :: [[\alpha]] \rightarrow [\alpha]
concat
           = foldr(#)[]
reverse :: [\alpha] \rightarrow [\alpha]
reverse = foldr snoc []
                 where snoc x xs = xs + [x]
length :: [\alpha] \rightarrow Int
length = foldr oneplus 0
                 where oneplus x n = 1 + n
            :: [Int] \rightarrow Int
sum
            = foldr (+) 0
sum
            :: (\alpha \rightarrow \beta) \rightarrow [\alpha] \rightarrow [\beta]
тар
            = foldr (cons \cdot f)
map f
                 where cons x xs = x : xs
```

```
def foldr[A,B](f: A => B => B)(e: B)(xs: List[A]): B =
 xs match {
   case Nil => e
   case x::xs => f(x)(foldr(f)(e)(xs))
  }
def concatenate[A]: List[A] => List[A] => List[A] =
 xs => ys => xs match {
   case Nil => ys
   case x :: xs => x :: concatenate(xs)(ys)
  }
def concat[A]: List[List[A]] => List[A] =
 foldr(concatenate[A])(Nil)
def reverse[A]: List[A] => List[A] = {
 def snoc[A]: A => List[A] => List[A] =
   x => xs => concatenate(xs)(List(x))
 foldr(snoc[A])(Nil)
def length[A]: List[A] => Int = {
 def oneplus[A]: A => Int => Int = x => n => 1 + n
 foldr(oneplus)(0)
val sum: List[Int] => Int = {
 val plus: Int => Int => Int = a => b => a + b
 foldr(plus)(0)
def map[A,B]: (A => B) => List[A] => List[B] = {
 def cons: B => List[B] => List[B] = x => xs => x :: xs
 f => foldr(cons compose f)(Nil)
```



```
assert( concatenate(List(1,2,3))(List(4,5)) == List(1,2,3,4,5) )
assert( concat(List(List(1,2), List(3), List(4,5))) == List(1,2,3,4,5) )
assert( reverse(List(1,2,3,4,5)) == List(5,4,3,2,1) )
assert( length(List(0,1,2,3,4,5)) == 6 )
assert( sum(List(2,3,4)) == 9 )
val mult: Int => Int => Int = a => b => a * b
assert( map(mult(10))(List(1,2,3)) == List(10,20,30))
```

It turns out that if it is possible to define a function on lists both using a **recursive definition** and using a definition in terms of *foldr*, then **there is a technique that can be used to go from the recursive definition to the definition using** *foldr*.



I came across the technique in the following paper by the author of **Programming in Haskell**:

A tutorial on the universality and expressiveness of fold

**GRAHAM HUTTON** 

University of Nottingham, Nottingham, UK http://www.cs.nott.ac.uk/-gmh

The tutorial (which I shall be referring to as **TUEF**), shows how to apply the technique to the **sum** function and the **map** function, which is the subject of the next five slides. **Note**: in the paper, the *foldr* function is referred to as *fold*.

A tutorial on the universality and expressiveness of fold

**GRAHAM HUTTON** 



Graham Hutton @haskellhutt

#### 3 The universal property of fold

As with the **fold** operator itself, the **universal property of** *fold* also has its origins in recursion theory. The first systematic use of the **universal property** in functional programming was by Malcolm (1990a), in his generalisation of Bird and Meerten's theory of lists (Bird, 1989; Meertens, 1983) to arbitrary regular datatypes. For finite lists, the universal property of *fold* can be stated as the following equivalence between two definitions for a function *g* that processes lists:

$$g [] = v \qquad \Leftrightarrow \qquad g = fold f v$$
$$g(x : xs) = f x (g xs)$$

In the right-to-left direction, substituting g = fold f v into the two equations for g gives the recursive definition for fold. Conversely, in the left-to-right direction the two equations for g are precisely the assumptions required to show that g = fold f v using a simple proof by induction on finite lists (Bird, 1998). Taken as a whole, the universal property states that for finite lists the function fold f v is not just a solution to its defining equations, but in fact the unique solution.... The universal property of fold can be generalised to handle partial and infinite lists (Bird, 1998), but for simplicity we only consider finite lists in this article.





That is, using the universal property we have calculated that:

sum = fold(+)0

Note that the key step ( $^{+}$ ) above in calculating a definition for f is the generalisation of the expression sum xs to a fresh variable y. In fact, such a generalisation step is not specific to the sum function, but will be a key step in the transformation of any recursive function into a definition using *fold* in this manner.

A tutorial on the universality and expressiveness of fold

GRAHAM HUTTON



Graham Hutton @haskellhutt

Of course, the *sum* example above is rather artificial, because the definition of *sum* using *fold* is immediate. However, there are many examples of functions whose definition using *fold* is not so immediate. For example, consider the recursively defined function *map* f that applies a function f to each element of a list:

 $\begin{array}{ll} map & :: (\alpha \to \beta) \to ([\alpha] \to [\beta]) \\ map f & [ ] & = [ ] \\ map f(x : xs) & = f x : map f xs \end{array}$ 

To redefine *map* f using *fold* we must solve the equation *map* f = fold v g for a function g and a value v. By appealing to the **universal property**, we conclude that this equation is equivalent to the following two equations:





That is, using the **universal property** we have calculated that

$$map f = fold (\lambda x ys \rightarrow f x : ys) []$$

In general, any function on lists that can be expressed using the fold operator can be transformed into such a definition using the universal property of fold.

Graham Hutton @haskellhutt



There are several other interesting things in **TUEF** that we'll be looking at.

I like its description of *foldr* (see right), because it reiterates a key point (see left) made by **Richard Bird** about **recursive functions** on lists.

Consider the following definition of a function *h* :

 $\begin{array}{l}h\left[\right] &= e\\h\left(x:xs\right) &= x \bigoplus h\,xs\end{array}$ 

```
The function h works by <u>taking a list</u>, <u>replacing</u> [] <u>by</u> e <u>and</u> (:) <u>by</u> \bigoplus, <u>and evaluating</u> <u>the result</u>. For example, h converts the list
```

```
x_1:(x_2:(x_3:(x_4:[])))
```

to the value

$$x_1 \oplus (x_2 \oplus (x_3 \oplus (x_4 \oplus e)))$$

Since (:) associates to the right, there is no need to put in parentheses in the first expression. However, we do need to put in parentheses in the second expression because we do not assume that  $\bigoplus$  associates to the right.

The pattern of definition given by h is captured in a function foldr (pronounced 'fold right') defined as follows:

$$\begin{array}{ll} foldr & :: (\alpha \rightarrow \beta \rightarrow \beta) \rightarrow \beta \rightarrow [\alpha] \rightarrow \beta \\ foldr \ f \ e \ [ \ ] & = e \\ foldr \ f \ e \ (x: xs) & = f \ x \ (foldr \ f \ e \ xs) \end{array}$$

A tutorial on the universality and expressiveness of fold

GRAHAM HUTTON

#### 2 The fold operator

The **fold** operator has its origins in recursion theory (Kleene, 1952), while the use of **fold** as a central concept in a programming language dates back to the reduction operator of APL (Iverson, 1962), and later to the insertion operator of FP (Backus, 1978). In **Haskell**, the **fold** operator for lists can be defined as follows:

$$\begin{array}{ll} fold & :: (\alpha \rightarrow \beta \rightarrow \beta) \rightarrow \beta \rightarrow ([\alpha] \rightarrow \beta) \\ fold \ f \ v \ [ \ ] & = v \\ fold \ f \ v \ (x : xs) & = f \ x \ (fold \ f \ v \ xs) \end{array}$$

That is, given a function f of type  $\alpha \rightarrow \beta \rightarrow \beta$  and a value v of type  $\beta$ , <u>the function</u> fold f v processes a list of type  $[\alpha]$  to give a value of type  $\beta$  by replacing the <u>nil</u> <u>constructor</u> [] at the end of the list by the value v, and each <u>cons constructor</u> (:) within the list by the function f. In this manner, the *fold* operator encapsulates a simple pattern of recursion for processing lists, in which the <u>two constructors for lists are simply</u> <u>replaced by other values and functions</u>.





Remember the **list concatenation** function we saw earlier?

$$(\texttt{#}) \qquad :: \quad [\alpha] \to [\alpha] \to [\alpha]$$
$$[] \texttt{#} ys \qquad = \quad ys$$
$$(x:xs) \texttt{#} ys \qquad = \quad x:(xs \texttt{#} ys)$$

Concatenation takes two lists, both of the same type, and produces a third list, again of the same type. def concatenate[A]: List[A] => List[A] => List[A] =
 xs => ys => xs match {
 case Nil => ys
 case x :: xs => x :: concatenate(xs)(ys)
 }

assert( concatenate(List(1,2,3))(List(4,5)) == List(1,2,3,4,5) )



In **TUEF** we find a definition of **concatenation** in terms of *foldr* (which it calls *fold*)

(#) ::	$[\alpha] \to [\alpha] \to [\alpha]$
(# ys) =	fold (:) ys

def concatenate[A]: List[A] => List[A] => List[A] = {
 def cons: A => List[A] => List[A] =
 x => xs => x :: xs
 xs => ys => foldr(cons)(ys)(xs)
}



Remember the *filter* function we saw earlier?

filter	::	$(\alpha \rightarrow Bool) \rightarrow [\alpha] \rightarrow [\alpha]$
<i>filter</i> p []	=	[]
<i>filter</i> $p(x:xs)$	=	<b>if</b> <i>p x</i> <b>then</b> <i>x</i> : <i>filter p xs</i> <b>else</b> <i>filter p xs</i>

def filter[A]: (A => Boolean) => List[A] => List[A] = p => {
 case Nil => Nil
 case x :: xs => if (p(x)) x :: filter(p)(xs) else filter(p)(xs)

val gt: Int => Int => Boolean = x => y => y > x
assert(filter(gt(5))(List(10,2,8,5,3,6)) == List(10,8,6))



In **TUEF** we find a definition of *filter* in terms of *foldr* (which as we saw, it calls *fold*)

filter	::	$(\alpha \rightarrow Bool) \rightarrow [\alpha] \rightarrow [\alpha]$
filter p	=	<i>fold</i> ( $\lambda x xs \rightarrow if p x$ then $x : xs$ else $xs$ ) []

def filter[A]: (A => Boolean) => List[A] => List[A] = p =>
foldr((x:A) => (xs:List[A]) => if (p(x)) (x::xs) else xs)(Nil)

Not every function on lists can be defined as an instance of *foldr*. For example, zip cannot be so defined. Even for those that can, an alternative definition may be more efficient. To illustrate, suppose we want a function *decimal* that takes a list of digits and returns the corresponding decimal number; thus

*decimal*  $[x_0, x_1, ..., x_n] = \sum_{k=0}^n x_k 10^{(n-k)}$ 

It is assumed that the most significant digit comes first in the list. One way to compute *decimal* efficiently is by a process of multiplying each digit by ten and adding in the following digit. For example

*decimal*  $[x_0, x_1, x_2] = 10 \times (10 \times (10 \times 0 + x_0) + x_1) + x_2$ 

This decomposition of a sum of powers is known as *Horner's* rule.

Suppose we define  $\bigoplus$  by  $n \bigoplus x = 10 \times n + x$ . Then we can rephrase the above equation as

*decimal*  $[x_0, x_1, x_2] = ((0 \oplus x_0) \oplus x_1) \oplus x_2$ 

This is almost like an instance of *foldr*, except that the grouping is the other way round, and the starting value appears on the left, not on the right. In fact the computation is dual: instead of processing from right to left, the computation processes from left to right.

This example motivates the introduction of a second fold operator called *foldl* (pronounced 'fold left'). Informally:

**foldl** ( $\oplus$ ) **e**  $[x_0, x_1, ..., xn_1] = (...((e \oplus x_0) \oplus x_1) ...) \oplus x_{n_1}$ 

The parentheses group from the left, which is the reason for the name. The full definition of *foldl* is

 $\begin{array}{ll} foldl & :: (\beta \rightarrow \alpha \rightarrow \beta) \rightarrow \beta \rightarrow [\alpha] \rightarrow \beta \\ foldl \ f \ e \ [ \ ] & = e \\ foldl \ f \ e \ (x: xs) = foldl \ f \ (f \ e \ x) \ xs \end{array}$ 





For example

 $\begin{array}{rcl} foldl (\bigoplus) & e & [x_0, x_1, x_2] \\ = & foldl (\bigoplus) & (e \oplus x_0) & [x_1, x_2] \\ = & foldl (\bigoplus) & ((e \oplus x_0) \oplus x_1) & [x_2] \end{array}$ 

 $= foldl (\oplus) (((e \oplus x_0) \oplus x_1) \oplus x_2) []$ 

$$= ((\mathbf{e} \oplus x_0) \oplus x_1) \oplus x_2$$

If  $\oplus$  is associative with unit *e*, then  $foldr(\oplus) e$  and  $foldl(\oplus) e$  define the same function on finite lists, as we will see in the following section.

As another example of the use of *foldl*, consider the following definition:

```
\begin{array}{ll} reverse' & :: & [\alpha] \rightarrow [\alpha] \\ reverse' & = & foldl \ cons \ [ \ ] \\ & where \ cons \ xs \ x = x \ : \ xs \end{array}
```

```
\begin{array}{rcl} reverse & :: & [\alpha] \rightarrow [\alpha] \\ reverse & = & foldr \ snoc \ [ \ ] \\ & where \ snoc \ x \ xs = xs \ \# \ [x] \end{array}
```

Note the order of the arguments to cons; we have cons = flip (:), where the standard function flip is defined by flipf x y = f y x. The function *reverse'*, reverses a finite list. For example:

 $reverse' [x_0, x_1, x_2] = cons(cons(cons[] x_0) x_1) x_2 = cons(cons[x_1] x_0) x_2 = cons[x_1, x_0] x_2 = [x_2, x_1, x_0]$ 

One can prove that reverse' = reverse by induction, or as an instance of a more general result in the following section. Of greater importance than the mere fact that *reverse* can be defined in a different way, is that *reverse'* gives a much more efficient program: reverse' takes time proportional to n on a list of length n, while reverse takes time proportional to  $n^2$ .







Here we can see the **Scala** version of *reverse*', and how it compares with *reverse* 

```
@philip_schwarz
```

```
(\texttt{#}) \quad :: \quad [\alpha] \to [\alpha] \to [\alpha] \\ (\texttt{#} ys) = \quad fold \ (:) \ ys
```

```
def concatenate[A]: List[A] => List[A] => List[A] = {
    def cons: A => List[A] => List[A] =
        x => xs => x :: xs
        xs => ys => foldr(cons)(ys)(xs)
}
```





That's it for part 1. I hope you enjoyed that.

There is still a lot to cover of course, so I'll see you in part 2.