

Functional Programming: Functional Programming

7. Folds, and Fold-Fusion

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1 Folds On Lists

A Common Pattern We've Seen Many Times...

$$\begin{aligned} \text{sum } [] &= 0 \\ \text{sum } (x : xs) &= x + \text{sum } xs \end{aligned}$$

$$\begin{aligned} \text{length } [] &= 0 \\ \text{length } (x : xs) &= 1 + \text{length } xs \end{aligned}$$

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

This pattern is extracted and called *foldr*:

$$\begin{aligned} \text{foldr } f e [] &= e, \\ \text{foldr } f e (x : xs) &= f x (\text{foldr } f e xs). \end{aligned}$$

For easy reference, we sometimes call e the “base value” and f the “step function.”

1.1 The Ubiquitous *foldr*

Replacing Constructors

$$\begin{aligned} \text{foldr } f e [] &= e \\ \text{foldr } f e (x : xs) &= f x (\text{foldr } f e xs) \end{aligned}$$

- One way to look at $\text{foldr } (\oplus) e$ is that it replaces $[]$ with e and $(:)$ with (\oplus) :

$$\begin{aligned} &\text{foldr } (\oplus) e [1, 2, 3, 4] \\ &= \text{foldr } (\oplus) e (1 : (2 : (3 : (4 : [])))) \\ &= 1 \oplus (2 \oplus (3 \oplus (4 \oplus e))). \end{aligned}$$

- $\text{sum} = \text{foldr } (+) 0$.
- $\text{length} = \text{foldr } (\lambda x n. 1 + n) 0$.
- $\text{map } f = \text{foldr } (\lambda x xs. f x : xs) []$.
- One can see that $\text{id} = \text{foldr } (:) []$.

Some Trivial Folds on Lists

- Function max returns the maximum element in a list:

$$\begin{aligned} \text{max } [] &= -\infty, \\ \text{max } (x : xs) &= x \uparrow \text{max } xs. \end{aligned}$$

$$\text{max} = \text{foldr } (\uparrow) -\infty.$$

- This function is actually called *maximum* in the standard Haskell Prelude, while max returns the maximum between its two arguments. For brevity, we denote the former by max and the latter by (\uparrow) .

- Function prod returns the product of a list:

$$\begin{aligned} \text{prod } [] &= 1, \\ \text{prod } (x : xs) &= x \times \text{prod } xs. \end{aligned}$$

$$\text{prod} = \text{foldr } (\times) 1.$$

- Function and returns the conjunction of a list:

$$\begin{aligned} \text{and } [] &= \text{true}, \\ \text{and } (x : xs) &= x \wedge \text{and } xs. \end{aligned}$$

$$\text{and} = \text{foldr } (\wedge) \text{ true}.$$

- Lets emphasise again that id on lists is a fold:

$$\begin{aligned} \text{id } [] &= [], \\ \text{id } (x : xs) &= x : \text{id } xs. \\ \text{id} &= \text{foldr } (:) []. \end{aligned}$$

Some Functions We Have Seen...

- $(++ ys) = \text{foldr } (:) ys$.

$$\begin{aligned} (++) &\quad :: \text{List } a \rightarrow \text{List } a \rightarrow \text{List } a \\ [] ++ ys &= ys \\ (x : xs) ++ ys &= x : (xs ++ ys) . \end{aligned}$$

- $\text{concat} = \text{foldr } (++) []$.

$$\begin{aligned} \text{concat} &\quad :: \text{List } (\text{List } a) \rightarrow \text{List } a \\ \text{concat } [] &= [] \\ \text{concat } (xs : xss) &= xs ++ \text{concat } xss . \end{aligned}$$

Replacing Constructors

- Understanding $foldr$ from its type. Recall

```
data List a = [] | a : List a .
```

- Types of the two constructors: $[] :: List a$, and $(:) :: a \rightarrow List a \rightarrow List a$.

- $foldr$ replaces the constructors:

$$\begin{aligned} foldr &:: (a \rightarrow b \rightarrow b) \rightarrow b \rightarrow List a \rightarrow b \\ foldr f e [] &= e \\ foldr f e (x : xs) &= f x (foldr f e xs) . \end{aligned}$$

Functions on Lists That Are Not $foldr$

- A function f is a $foldr$ if in $f (x : xs) = ...f xs...$, the argument xs does not appear outside of the recursive call.
- Not all functions taking a list as input is a $foldr$.
- The canonical example is perhaps $tail :: List a \rightarrow List a$.
 - $- tail(x : xs) = ...tail xs..??$
 - $- tail$ dropped too much information, which cannot be recovered.
- Another example is $dropWhile p :: List a \rightarrow List a$.

Longest Prefix

- The function call $takeWhile p xs$ returns the longest prefix of xs that satisfies p :

$$\begin{aligned} takeWhile p [] &= [] \\ takeWhile p (x : xs) &= \\ &\text{if } p x \text{ then } x : takeWhile p xs \\ &\text{else } [] . \end{aligned}$$

- E.g. $takeWhile (\leq 3) [1, 2, 3, 4, 5] = [1, 2, 3]$.
- It can be defined by a fold:

$$takeWhile p \\ foldr (\lambda x xs \rightarrow \text{if } p x \text{ then } x : xs \text{ else } []) [] .$$

All Prefixes

- The function $inits$ returns the list of all prefixes of the input list:

$$\begin{aligned} inits [] &= [[]], \\ inits (x : xs) &= [] : map (x :) (inits xs). \end{aligned}$$

- E.g. $inits [1, 2, 3] = [[], [1], [1, 2], [1, 2, 3]]$.

- It can be defined by a fold:

$$inits = foldr (\lambda x xss \rightarrow [] : map (x :) xss) [[]].$$

All Suffixes

- The function $tails$ returns the list of all suffixes of the input list:

$$\begin{aligned} tails [] &= [[]], \\ tails (x : xs) &= (x : xs) : tails xs. \end{aligned}$$

- It appears that $tails$ is not a $foldr$!

- Luckily, we have $head (tails xs) = xs$. Therefore,

$$\begin{aligned} tails (x : xs) &= \text{let } yss = tails xs \\ &\text{in } (x : head yss) : yss. \end{aligned}$$

- The function $tails$ may thus be defined by a fold:

$$tails = foldr (\lambda x yss \rightarrow \\ (x : head yss) : yss) [[]].$$

1.2 The Fold-Fusion Theorem

Why Folds?

- “What are the three most important factors in a programming language?” Abstraction, abstraction, and abstraction!
- Control abstraction, procedure abstraction, data abstraction,...can programming patterns be abstracted too?
- Program structure becomes an entity we can talk about, reason about, and reuse.
 - We can describe algorithms in terms of fold, unfold, and other recognised patterns.
 - We can prove properties about folds,
 - and apply the proved theorems to all programs that are folds, either for compiler optimisation, or for mathematical reasoning.
- Among the theorems about folds, the most important is probably the *fold-fusion* theorem.

The Fold-Fusion Theorem

The theorem is about when the composition of a function and a fold can be expressed as a fold.

Theorem 1 (foldr-Fusion). Given $f :: a \rightarrow b \rightarrow b$, $e :: b$, $h :: b \rightarrow c$, and $g :: a \rightarrow c \rightarrow c$, we have:

$$h \cdot \text{foldr } f e = \text{foldr } g (h e) ,$$

if $h(f x y) = g x (h y)$ for all x and y .

For program derivation, we are usually given h , f , and e , from which we have to construct g .

Tracing an Example

Let us try to get an intuitive understand of the theorem:

$$\begin{aligned} & h(\text{foldr } f e [a, b, c]) \\ &= \{ \text{definition of foldr} \} \\ & h(f a (f b (f c e))) \\ &= \{ \text{since } h(f x y) = g x (h y) \} \\ & g a (h(f b (f c e))) \\ &= \{ \text{since } h(f x y) = g x (h y) \} \\ & g a (g b (h(f c e))) \\ &= \{ \text{since } h(f x y) = g x (h y) \} \\ & g a (g b (g c (h e))) \\ &= \{ \text{definition of foldr} \} \\ & \text{foldr } g (h e) [a, b, c] . \end{aligned}$$

Sum of Squares, Again

- Consider $\text{sum} \cdot \text{map square}$ again. This time we use the fact that $\text{map } f = \text{foldr } (\text{mf } f) []$, where $\text{mf } f x xs = f x : xs$.
- $\text{sum} \cdot \text{map square}$ is a fold, if we can find a ssq such that $\text{sum}(\text{mf square } x xs) = ssq x (\text{sum } xs)$. Let us try:

$$\begin{aligned} & \text{sum}(\text{mf square } x xs) \\ &= \{ \text{definition of mf} \} \\ & \text{sum}(\text{square } x : xs) \\ &= \{ \text{definition of sum} \} \\ & \text{square } x + \text{sum } xs \\ &= \{ \text{let } ssq x y = \text{square } x + y \} \\ & ssq x (\text{sum } xs) . \end{aligned}$$

Therefore, $\text{sum} \cdot \text{map square} = \text{foldr } ssq 0$.

Sum of Squares, without Folds

Recall that this is how we derived the inductive case of sumsq yesterday:

$$\begin{aligned} & \text{sumsq } (x : xs) \\ &= \{ \text{definition of sumsq} \} \\ & \text{sum}(\text{map square } (x : xs)) \\ &= \{ \text{definition of map} \} \\ & \text{sum}(\text{square } x : \text{map square } xs) \\ &= \{ \text{definition of sum} \} \\ & \text{square } x + \text{sum}(\text{map square } xs) \\ &= \{ \text{definition of sumsq} \} \\ & \text{square } x + \text{sumsq } xs . \end{aligned}$$

Comparing the two derivations, by using fold-fusion we supply only the “important” part.

More on Folds and Fold-fusion

- Compare the proof with the one yesterday. They are essentially the same proof.
- Fold-fusion theorem abstracts away the common parts in this kind of inductive proofs, so that we need to supply only the “important” parts.

Scan

- The following function scanr computes foldr for every suffix of the given list:

$$\begin{aligned} \text{scanr} &:: (a \rightarrow b \rightarrow b) \rightarrow b \rightarrow \text{List } a \rightarrow \text{List } b \\ \text{scanr } f e &= \text{map}(\text{foldr } f e) \cdot \text{tails} . \end{aligned}$$

- E.g. computing the running sum of a list:

$$\begin{aligned} & \text{scanr } (+) 0 [8, 1, 3] \\ &= \text{map } \text{sum}(\text{tails } [8, 1, 3]) \\ &= \text{map } \text{sum}[[8, 1, 3], [1, 3], [3], []] \\ &= [12, 4, 3, 0]. \end{aligned}$$

- Surely there is a quicker way to compute scanr , right?

Scan

- Recall that tails is a foldr :

$$\begin{aligned} \text{tails} &= \text{foldr } (\lambda x yss \rightarrow \\ & (x : \text{head } yss) : yss) [[]] . \end{aligned}$$

- By *foldr*-fusion we get:

$$\text{scanr } f e = \text{foldr} (\lambda x ys \rightarrow f x (\text{head } ys) : ys) [e],$$

- which is equivalent to this inductive definition:

$$\begin{aligned} \text{scanr } f e [] &= [e] \\ \text{scanr } f e (x : xs) &= f x (\text{head } ys) : ys, \\ &\text{where } ys = \text{scanr } f e xs. \end{aligned}$$

Tupling as Fold-fusion

- Tupling can be seen as a kind of fold-fusion. The derivation of *steepsum*, for example, can be seen as fusing:

$$\text{steepsum} \cdot id = \text{steepsum} \cdot \text{foldr} (:) [] .$$

- Recall that $\text{steepsum } xs = (\text{steep } xs, \text{sum } xs)$. Reformulating *steepsum* into a fold allows us to compute it in one traversal.

Accumulating Parameter as Fold-Fusion

- We also note that introducing an accumulating parameter can often be seen as fusing a higher-order function with a *foldr*.
- Recall the function *reverse*. Observe that

$$\text{reverse} = \text{foldr} (\lambda x xs \rightarrow xs ++ [x]) [] .$$

- Recall $\text{revcat } xs ys = \text{reverse } xs ++ ys$. It is equivalent to

$$\text{revcat} = (++) \cdot \text{reverse} .$$

- Deriving *revcat* is performing a fusion!

2 Folds on Other Algebraic Datatypes

- Folds are a specialised form of induction.
- Inductive datatypes: types on which you can perform induction.
- Every inductive datatype give rise to its fold.
- In fact, an inductive type can be defined by its fold.

Fold on Natural Numbers

- Recall the definition:

$$\text{data } Nat = 0 \mid \mathbf{1}_+ \text{Nat} .$$

- Constructors: $0 :: Nat$, $(\mathbf{1}_+) :: Nat \rightarrow Nat$.

- What is the fold on *Nat*?

$$\begin{aligned} \text{foldN} &:: (a \rightarrow a) \rightarrow a \rightarrow Nat \rightarrow a \\ \text{foldN } f e 0 &= e \\ \text{foldN } f e (\mathbf{1}_+ n) &= f (\text{foldN } f e n) . \end{aligned}$$

Examples of *foldN*

- $(+n) = \text{foldN } (\mathbf{1}_+) n$.

$$\begin{aligned} 0 + n &= n \\ (\mathbf{1}_+ m) + n &= \mathbf{1}_+ (m + n) . \end{aligned}$$

- $(\times n) = \text{foldN } (n+) 0$.

$$\begin{aligned} 0 \times n &= 0 \\ (\mathbf{1}_+ m) \times n &= n + (m \times n) . \end{aligned}$$

- $\text{even} = \text{foldN } \text{not } \text{True}$.

$$\begin{aligned} \text{even } 0 &= \text{True} \\ \text{even } (\mathbf{1}_+ n) &= \text{not } (\text{even } n) . \end{aligned}$$

Fold-Fusion for Natural Numbers

Theorem 2 (*foldN*-Fusion). Given $f :: a \rightarrow a$, $e :: a$, $h :: a \rightarrow b$, and $g :: b \rightarrow b$, we have:

$$h \cdot \text{foldN } f e = \text{foldN } g (h e) ,$$

if $h (f x) = g (h x)$ for all x .

Exercise: fuse *even* into $(+)$?

Folds on Trees

- Recall some datatypes for trees:

$$\begin{aligned} \text{data } ITree a &= \text{Null} \mid \text{Node } a (ITree a) (ITree a) , \\ \text{data } ETree a &= \text{Tip } a \mid \text{Bin } (ETree a) (ETree a) . \end{aligned}$$

- The fold for *ITree*, for example, is defined by:

$$\begin{aligned} \text{foldIT} &:: (a \rightarrow b \rightarrow b \rightarrow b) \rightarrow b \rightarrow ITree a \rightarrow b \\ \text{foldIT } f e \text{ Null} &= e \\ \text{foldIT } f e (\text{Node } a t u) &= f a (\text{foldIT } f e t) (\text{foldIT } f e u) . \end{aligned}$$

- The fold for *ETree*, is given by:

$$\begin{aligned} \text{foldET} &:: (b \rightarrow b \rightarrow b) \rightarrow (a \rightarrow b) \rightarrow ETree a \rightarrow b \\ \text{foldET } f k (\text{Tip } x) &= k x \\ \text{foldET } f k (\text{Bin } t u) &= f (f \text{oldET } f k t) (f \text{oldET } f k u) . \end{aligned}$$

Some Simple Functions on Trees

- To compute the size of an *ITree*:

$$\text{sizeITree} = \text{foldIT} (\lambda x m n \rightarrow \mathbf{1}_+ (m + n)) 0 .$$

- To sum up labels in an *ETree*:

$$\text{sumETree} = \text{foldET} (+) \text{id} .$$

- To compute a list of all labels in an *ITree* and an *ETree*:

$$\text{flattenIT} = \text{foldIT} (\lambda x xs ys \rightarrow xs ++ [x] ++ ys) [] , \quad \text{map sum} (\text{ini } x \text{xss})$$

$$\text{flattenET} = \text{foldET} (++) (\lambda x \rightarrow [x]) .$$

- **Exercise:** what are the fusion theorems for *foldIT* and *foldET*?

3 Finally, Solving Maximum Segment Sum

Specifying Maximum Segment Sum

- Finally we have introduced enough concepts to tackle the maximum segment sum problem!
- A segment can be seen as a prefix of a suffix.
- The function *segs* computes the list of all the segments.

$$\text{segs} = \text{concat} \cdot \text{map inits} \cdot \text{tails} .$$

- Therefore, *mss* is specified by:

$$\text{mss} = \text{max} \cdot \text{map sum} \cdot \text{segs} .$$

The Derivation!

We reason:

$$\begin{aligned} & \text{max} \cdot \text{map sum} \cdot \text{concat} \cdot \text{map inits} \cdot \text{tails} \\ &= \{ \text{since } \text{map } f \cdot \text{concat} = \text{concat} \cdot \text{map} (\text{map } f) \} \\ &\quad \text{max} \cdot \text{concat} \cdot \text{map} (\text{map sum}) \cdot \text{map inits} \cdot \text{tails} \\ &= \{ \text{since } \text{max} \cdot \text{concat} = \text{max} \cdot \text{map max} \} \\ &\quad \text{max} \cdot \text{map max} \cdot \text{map} (\text{map sum}) \cdot \text{map inits} \cdot \text{tails} \\ &= \{ \text{since } \text{map } f \cdot \text{map } g = \text{map} (f.g) \} \\ &\quad \text{max} \cdot \text{map} (\text{max} \cdot \text{map sum} \cdot \text{inits}) \cdot \text{tails} . \end{aligned}$$

Recall the definition $\text{scanr } f e = \text{map} (\text{foldr } f e) \cdot \text{tails}$. If we can transform $\text{max} \cdot \text{map sum} \cdot \text{inits}$ into a fold, we can turn the algorithm into a *scanr*, which has a faster implementation.

Maximum Prefix Sum

Concentrate on $\text{max} \cdot \text{map sum} \cdot \text{inits}$ ($\text{let } \text{ini } x \text{xss} = [] : \text{map} (x :) \text{xss} :$

$$\begin{aligned} &\text{max} \cdot \text{map sum} \cdot \text{inits} \\ &= \{ \text{definition of } \text{init}, \text{ini } x \text{xss} = [] : \text{map} (x :) \text{xss} \} \\ &\quad \text{max} \cdot \text{map sum} \cdot \text{foldr } \text{ini} [] \\ &= \{ \text{fold fusion, see below} \} \\ &\quad \text{max} \cdot \text{foldr } \text{zplus} [0] . \end{aligned}$$

The fold fusion works because:

$$\begin{aligned} &\text{flattenIT} = \text{foldIT} (\lambda x xs ys \rightarrow xs ++ [x] ++ ys) [] , \quad \text{map sum} (\text{ini } x \text{xss}) \\ &= \text{map sum} ([] : \text{map} (x :) \text{xss}) \\ &= 0 : \text{map} (\text{sum} \cdot (x :)) \text{xss} \\ &= 0 : \text{map} (x+) (\text{map sum} \text{xss}) . \end{aligned}$$

Define $\text{zplus } x \text{yss} = 0 : \text{map} (x+) \text{yss}$.

Maximum Prefix Sum, 2nd Fold Fusion

Concentrate on $\text{max} \cdot \text{map sum} \cdot \text{inits}$:

$$\begin{aligned} &\text{max} \cdot \text{map sum} \cdot \text{inits} \\ &= \{ \text{definition of } \text{init}, \text{ini } x \text{xss} = [] : \text{map} (x :) \text{xss} \} \\ &\quad \text{max} \cdot \text{map sum} \cdot \text{foldr } \text{ini} [] \\ &= \{ \text{fold fusion, } \text{zplus } x \text{xss} = 0 : \text{map} (x+) \text{xss} \} \\ &\quad \text{max} \cdot \text{foldr } \text{zplus} [0] \\ &= \{ \text{fold fusion, let } \text{zmax } x y = 0 \uparrow (x + y) \} \\ &\quad \text{foldr } \text{zmax} 0 . \end{aligned}$$

The fold fusion works because \uparrow distributes into $(+)$:

$$\begin{aligned} &\text{max} (0 : \text{map} (x+) \text{xs}) \\ &= 0 \uparrow \text{max} (\text{map} (x+) \text{xs}) \\ &= 0 \uparrow (x + \text{max} \text{xs}) . \end{aligned}$$

Back to Maximum Segment Sum

We reason:

$$\begin{aligned} &\text{max} \cdot \text{map sum} \cdot \text{concat} \cdot \text{map inits} \cdot \text{tails} \\ &= \{ \text{since } \text{map } f \cdot \text{concat} = \text{concat} \cdot \text{map} (\text{map } f) \} \\ &\quad \text{max} \cdot \text{concat} \cdot \text{map} (\text{map sum}) \cdot \text{map inits} \cdot \text{tails} \\ &= \{ \text{since } \text{max} \cdot \text{concat} = \text{max} \cdot \text{map max} \} \\ &\quad \text{max} \cdot \text{map max} \cdot \text{map} (\text{map sum}) \cdot \text{map inits} \cdot \text{tails} \\ &= \{ \text{since } \text{map } f \cdot \text{map } g = \text{map} (f.g) \} \\ &\quad \text{max} \cdot \text{map} (\text{max} \cdot \text{map sum} \cdot \text{inits}) \cdot \text{tails} \\ &= \{ \text{reasoning in the previous slides} \} \\ &\quad \text{max} \cdot \text{map} (\text{foldr } \text{zmax} 0) \cdot \text{tails} \\ &= \{ \text{introducing } \text{scanr} \} \\ &\quad \text{max} \cdot \text{scanr } \text{zmax} 0 . \end{aligned}$$

Maximum Segment Sum in Linear Time!

- We have derived $mss = max \cdot scanr zmax 0$, where $zmax x y = 0 \uparrow (x + y)$.
- The algorithm runs in linear time, but takes linear space.
- A tupling transformation eliminates the need for linear space.

$$mss = fst \cdot maxhd \cdot scanr zmax 0$$

where $maxhd xs = (max xs, head xs)$. We omit this last step in the lecture.

- The final program is $mss = fst \cdot foldr step (0, 0)$, where $step x (m, y) = ((0 \uparrow (x + y)) \uparrow m, 0 \uparrow (x + y))$.