Programming Languages: Functional Programming 4. Simple Program Calculation

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A Quick Review

- Functions are the basic building blocks. They may be passed as arguments, may return functions, and can be composed together.
- While one issues commands in an imperative language, in functional programming we specify values, and computers try to reduce the values to their normal forms.
- Formal reasoning: reasoning with the form (syntax) rather than the semantics. Let the symbols do the work!
- 'Wholemeal' programming: think of aggregate data as a whole, and process them as a whole.
- Once you describe the values as algebraic datatypes, most programs write themselves through structural recursion.
- Programs and their proofs are closely related. They share similar structure, by induction over input data.
- Properties of programs can be reasoned about in equations, just like high school algebra.

1 Some Comments on Efficiency

Data Representation

- So far we have (surprisingly) been talking about mathematics without much concern regarding efficiency. Time for a change.
- Take lists for example. Recall the definition: data List a = [] | a : List a.
- Our representation of lists is biased. The left most element can be fetched immediately.

- Thus. (:), *head*, and *tail* are constant-time operations, while *init* and *last* takes linear-time.
- In most implementations, the list is represented as a linked-list.

List Concatenation Takes Linear Time

• Recall (++):

[] ++ ys = ys(x : xs) ++ ys = x : (xs ++ ys)

• Consider [1, 2, 3] + [4, 5]:

(1:2:3:[]) ++(4:5:[]) = 1:((2:3:[]) ++(4:5:[])) = 1:2:((3:[]) ++(4:5:[])) = 1:2:3:([] ++(4:5:[])) = 1:2:3:4:5:[])

• (++) runs in time proportional to the length of its left argument.

Full Persistency

- Compound data structures, like simple values, are just values, and thus must be *fully persistent*.
- That is, in the following code:

let xs = [1, 2, 3] ys = [4, 5] zs = xs ++ ysin ... body ...

• The *body* may have access to all three values. Thus ++ cannot perform a destructive update.

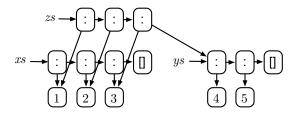


Figure 1: How (++) allocates new (:) cells in the heap.

Linked v.s. Block Data Structures

- Trees are usually represented in a similar manner, through links.
- Fully persistency is easier to achieve for such linked data structures.
- Accessing arbitrary elements, however, usually takes linear time.
- In imperative languages, constant-time random access is usually achieved by allocating lists (usually called arrays in this case) in a consecutive block of memory.
- Consider the following code, where *xs* is an array (implemented as a block), and *ys* is like *xs*, apart from its 10th element:

let xs = [1..100] $ys = update \ xs \ 10 \ 20$ in ... body ...

- To allow access to both *xs* and *ys* in *body*, the *update* operation has to duplicate the entire array.
- Thus people have invented some smart data structure to do so, in around $O(\log n)$ time.
- On the other hand, *update* may simply overwrite *xs* if we can somehow make sure that *nobody* other than *ys* uses *xs*.
- Both are advanced topics, however.

Another Linear-Time Operation

• Taking all but the last element of a list:

 $\begin{array}{l} init \ [x] &= [\] \\ init \ (x:xs) \ = x: init \ xs \end{array}$

• Consider *init* [1, 2, 3, 4]:

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init (1:2:3:4:[]) = 1:init (2:3:4:[]) = 1:2:init (3:4:[]) = 1:2:3:init (3:4:[]) = 1:2:3:init (4:[]) = 1:2:3:[]
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Sum, Map, etc

- Functions like *sum*, *maximum*, etc. needs to traverse through the list once to produce a result. So their running time is definitely O(n), where n is the length of the list.
- If f takes time O(t), map f takes time $O(n \times t)$ to complete. Similarly with filter p.
 - In a lazy setting, map f produces its first result in O(t) time. We won't need lazy features for now, however.

2 A First Taste of Program Calculation

Sum of Squares

- Given a sequence a_1, a_2, \dots, a_n , compute $a_1^2 + a_2^2 + \dots + a_n^2$. Specification: $sumsq = sum \cdot map \ square$.
- The spec. builds an intermediate list. Can we eliminate it?
- The input is either empty or not. When it is empty:

sumsq[]

- = { definition of sumsq }
 (sum · map square) []
 = { function composition }
 sum (map square [])
- $= \{ \text{ definition of } map \}$
 - sum []
- $= \{ \begin{array}{l} \text{definition of } sum \\ 0 \end{array} \}$

Sum of Squares, the Inductive Case

• Consider the case when the input is not empty:

- $= \{ \text{ definition of } sumsq \}$
- $sum (map \ square \ (x : xs))$
- = { definition of map }
 sum (square x : map square xs)
 = { definition of sum }
- square x + sum (map square xs)
- $= \{ \text{ definition of } sumsq \}$
 - $square \ x + sumsq \ xs$

Alternative Definition for *sumsq*

• From $sumsq = sum \cdot map \ square$, we have proved that

sumsq [] = 0sumsq (x : xs) = square x + sumsq xs

- Equivalently, we have shown that $sum \cdot map \ square$ is a solution of

$$f[] = 0$$

$$f(x:xs) = square x + f xs$$

- However, the solution of the equations above is unique.
- Thus we can take it as another definition of *sumsq*. Denotationally it is the same function; operationally, it is (slightly) quicker.
- Exercise: try calculating an inductive definition of *count*.

How Far Can We Get?

• Specification of maximum segment sum:

- Or, segments $xs = [zs \mid ys \leftarrow tails \ xs, zs \leftarrow inits \ ys].$
- From the specification we can calculate a linear time algorithm.

Remark: Why Functional Programming?

- Time to muse on the merits of functional programming. Why functional programming?
 - Algebraic datatype? List comprehension? Lazy evaluation? Garbage collection? These are just language features that can be migrated.
 - No side effects.¹ But why taking away a language feature?
- By being pure, we have a simpler semantics in which we are allowed to construct and reason about programs.
 - In an imperative language we do not even have $f 4 + f 4 = 2 \times f 4$.
- Ease of reasoning. That's the main benefit we get.

¹Unless introduced in a disciplined way. See Section ??.