Recitation 3 — Scan
Parallel and Sequential Data Structures and Algorithms, 15-210 (Spring 2013)

January 30, 2013

1 Announcements

- HW 2 is due next Monday. Hopefully you have all started by now; if not, now would be a good
time.
- Questions from lecture or homework?

2 Scan

Yesterday, we covered the function \texttt{scan}. We’ll recap the definition of \texttt{scan} briefly today, and show
you how to solve interesting problems with it.

\texttt{scan} takes a function as one of its arguments. All of the text below makes the assumption that this
function is \textit{associative}. Recall the mathematical definition that a function \( f \) is said to be associative if
and only if

\[
\forall a \forall b \forall c. f(f(a, b), c) = f(a, f(b, c))
\]

We also make the assumption that the initial value is a \textit{left-identity} of the functional argument. Recall
the mathematical definition that \( I \) is a left-identity of \( f \) if and only if

\[
\forall a. f(I, a) = a
\]

We don’t need these assumptions in general, and we’ll come back to a version of \texttt{scan} later that
doesn’t have them, but it’s a cleaner way to start thinking about \texttt{scan} with these properties.

With the assumption that \( f \) is associative, \((\texttt{scan} f b)\) is logically equivalent to \((\texttt{iterh} f b)\) in
the same way that \((\texttt{reduce} f b)\) is logically equivalent to \((\texttt{iter} f b)\).

Specifically, if \( f \) is a function that takes no more than a constant number of steps on all input, \((\texttt{iterh} f)\) and \((\texttt{iter} f)\) have both work and span in \( O(n) \), whereas \texttt{reduce} and \texttt{scan} both have work in
\( O(n) \) and span in \( O(\lg n) \).

It's worth noting that while \texttt{reduce} and \texttt{scan} are highly parallel, unlike \texttt{iter} and \texttt{iterh}, they pay
the price by having slightly less general types.

2.1 Note on Terminology

If \( f \) is a function and \( I \) is a relevant identity for \( f \), we’ll often say “\texttt{f-scan}” to mean
\[
\text{fn } s \Rightarrow \text{scan } f \ I
\]

For example, a “+-scan” is
\[
\text{fn } s \Rightarrow \text{scan } \left(\text{op +}\right) 0
\]

2.2 Recap

If \( s = \langle 1, 6, 3, -2, 9, 0, -4 \rangle \), then
\[
(\text{scan Int.min Int.maxInt } s)
\]
yields the following:
\[
(\langle \text{Int.maxInt}, 1, 1, 1, -2, -2, -2 \rangle, -4)
\]
Remember that in the result, location \( i \) stores the “sum” of the values at locations before \( i \) in the original sequence. There is a variant of \text{scan} called \text{scanI} which sums the values at locations before and including \( i \).

2.3 Example Uses of Scan

At first glance, \text{scan} seems to offer not much that isn’t already available through \text{reduce}. With clever choices of associative functions, though, \text{scan} can be used to compute some surprising things efficiently in parallel.

2.3.1 Histogram

Consider the following problem:

Given a histogram, if we were to pour water over it, how much water (in terms of area) would it hold? For simplicity we will represent a histogram as a sequence of non-negative integers. For example the histogram shown below is represented by the sequence \( s = (2, 3, 4, 7, 5, 2, 3, 2, 6, 4, 3, 5, 2, 1) \), and holds 15 units of water.
Any ideas on how we might solve this problem?

The idea is to single out one bar \( b_i \). If we know the maximum of the bar heights to the left of \( b_i \) (\( \text{max}_L \)) and the maximum of the bar heights to the right of \( b_i \) (\( \text{max}_R \)), given that \( \text{max}_L > \text{height}(b_i) \) and \( \text{max}_R > \text{height}(b_i) \) then the water \( b_i \) will hold above it is \( \min(\text{max}_L, \text{max}_R) - \text{height}(b_i) \).

Do we know of any functions that could be useful for generating these sequences of max-bar-heights? How about \text{scan}. Using a few \text{scan}'s, a \text{map} and a \text{reduce}, this problem becomes very simple.

```haskell
fun rev s = 
    let val n = length s 
        in tabulate (fn i => nth s (n - i - 1)) n 
    end

fun histogramFill (hist : int seq) = 
    let 
        val (lHeights, _) = scan Int.max 0 hist 
        val (rHeightsRev, _) = scan Int.max 0 (rev hist) 
        val heights = map2 Int.min lHeights (rev rHeightsRev) 

        fun nonNegative (maxHeight, thisHeight) = 
            Int.max (maxHeight - thisHeight, 0) 
        in 
            reduce op+ 0 (map2 nonNegative heights hist) 
        end

To get the maximum height to the left of each position, we use a \text{scan} with the \text{Int.max} operator. How about getting the maximum height to the right of each position? Use a \text{scan} on the reverse of the list, and then reverse it again! We can then compute the total fill with a \text{map} and a \text{reduce}.

### 2.3.2 Computing Fibonacci Numbers

With a carefully chosen matrix, we can use \text{scan} to compute the Fibonacci numbers. In the extremely unlikely event that you've forgotten, the Fibonacci numbers are defined as follows:

**Definition:** The Fibonacci numbers are an integer sequence given by the following recurrence\(^1\)

\[
\begin{align*}
F_{-1} &= 1 \\
F_0 &= 0 \\
F_1 &= 1 \\
F_n &= F_{n-1} + F_{n-2}
\end{align*}
\]

\(^1\)It is slightly contrived, but harmless, to define the \( -1^{\text{st}} \) element of the Fibonacci sequence. The other base cases are such that the recursive case will never use it, so this could be any constant and produce the same sequence of integers. This one happens to make the proof work, though.
We make the following claim about this definition, which we will prove by induction:

**Claim:**

For all natural numbers \( n \),

\[
\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^n = \begin{pmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{pmatrix}
\]

**Proof:** We’ll prove this by induction on \( n \).

**Base Case:** \( n = 0 \)

Any \( n \times n \) matrix to the zero power is the \( n \times n \) identity matrix, so

\[
\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^0 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} F_1 & F_0 \\ F_0 & F_{-1} \end{pmatrix}
\]

which is exactly as desired.

**Inductive Case:**

Assume that

\[
\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^n = \begin{pmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{pmatrix}
\]

We want to show that

\[
\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^{n+1} = \begin{pmatrix} F_{n+2} & F_{n+1} \\ F_{n+1} & F_n \end{pmatrix}
\]

It suffices to show that

\[
\begin{pmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{pmatrix} \cdot \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} F_{n+2} & F_{n+1} \\ F_{n+1} & F_n \end{pmatrix}
\]

Recall matrix multiplication, specifically in the case of taking the product of two \( 2 \times 2 \) matrices:

\[
\begin{pmatrix} a & b \\ c & d \end{pmatrix} \cdot \begin{pmatrix} e & f \\ g & h \end{pmatrix} = \begin{pmatrix} ae + bg & af + bh \\ ce + dg & cf + dh \end{pmatrix}
\]

Therefore,

\[
\begin{pmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{pmatrix} \cdot \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} F_{n+1} + F_n & F_{n+1} \\ F_n + F_{n-1} & F_n \end{pmatrix}
\]

\[
= \begin{pmatrix} F_{n+2} & F_{n+1} \\ F_{n+1} & F_n \end{pmatrix}
\]

This is exactly as desired and concludes the proof.
Remember that matrix multiplication is an associative operation on square matrices. We'll only need $2 \times 2$ int matrices, so for simplicity let's represent them as values of type `int * int * int * int`.

The above proof means that we can compute the Fibonacci numbers by applying scan to a matrix multiplication function:

(* very simple representation of 2x2 matrices *)

```ml
fun mmult ((a,b,c,d),(e,f,g,h)) = (a*e + b*g, a*f + b*h,
                                      c*e + d*g, c*f + d*h)
```

(* returns the first n fibonacci numbers *)

```ml
fun fib n = 
  let
    val s = tabulate (fn _ => (1,1,1,0)) n
    val (ans,_) = scan mmult (1,0,0,1) s
  in
    map (fn (_,x,_,_) => x) ans
  end
```

Since the matrices are of a constant $2 \times 2$ size, the matrix multiplication is just the cost of a constant number of integer additions and multiplications. If we do not include the complexity of additions and multiplications, then the work is $O(n)$ and span is $O(\log n)$, as we are just using `scan`. You will explore the complexity of integer addition and multiplication in HW 3.

### 2.3.3 Matching Parentheses

We can use `scan` to solve the parenthesis matching problem that we went over two weeks ago. The idea is that we first map each open parenthesis to 1 and each close parenthesis to $-1$. We then do a `+.-scan` on this integer sequence. The elements in the sequence returned by `scan` exactly correspond how many unmatched parentheses there are in that prefix of the string.

For example:

```
((,),((,),,))
```

becomes

```
(1, -1, 1, 1, -1, -1, -1)
```

and then

```
(0, 1, 0, 1, 2, 1, 0, -1)
```

and then fails, because the counter went negative at some point indicating an imbalance.

```ml
fun match s = 
  let
    fun paren2int OPAREN = 1
        | paren2int CPAREN = ~1
```

val C = map paren2int s
val (S,total) = scan (op+) 0 C
val SOME(maxint) = Int.maxInt

in
  (reduce Int.min maxint S) >= 0 andalso total = 0
end