# Parallel Prefix Scan and Assignment 7 <br> <br> COS 326 <br> <br> COS 326 <br> <br> Andrew Appel 

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## The prefix-sum problem

prefix_sum : int seq -> int seq


The simple sequential algorithm: accumulate the sum from left to right

- Sequential algorithm: Work: $O(n)$, Span: $O(n)$
- Goal: a parallel algorithm with Work: $O(n)$, Span: $O(\log n)$


## Parallel prefix-sum

The trick: Use two passes

- Each pass has $O(n)$ work and $O(\log n)$ span
- So in total there is $O(n)$ work and $O(\log n)$ span

First pass builds a tree of sums bottom-up

- the "up" pass

Second pass traverses the tree top-down to compute prefixes

- the "down" pass computes the "from-left-of-me" sum

Historical note:

- Original algorithm due to R. Ladner and M. Fischer, 1977




## The algorithm, pass 1

1. Up: Build a binary tree where

- Root has sum of the range $[\mathbf{x}, \mathbf{y})$
- If a node has sum of [lo,hi) and hi>lo,
- Left child has sum of [lo,middle)
- Right child has sum of [middle, hi)
- A leaf has sum of [i,i+1), i.e., nth input i

This is an easy parallel divide-and-conquer algorithm: "combine" results by actually building a binary tree with all the range-sums

- Tree built bottom-up in parallel

Analysis: $O(n)$ work, $O(\log n)$ span

## The algorithm, pass 2

2. Down: Pass down a value fromLeft

- Root given a fromLeft of 0
- Node takes its fromLeft value and
- Passes its left child the same fromLeft
- Passes its right child its fromLeft plus its left child's sum
- as stored in part 1
- At the leaf for sequence position $i$,
- nth output i == fromLeft + nth input i

This is an easy parallel divide-and-conquer algorithm:
traverse the tree built in step 1 and produce no result

- Leaves create output
- Invariant: fromLeft is sum of elements left of the node's range

Analysis: $O(n)$ work, $O(\log n)$ span

## Sequential cut-off

For performance, we need a sequential cut-off:

- Up:
- just a sum, have leaf node hold the sum of a range
- Down:
- do a sequential scan


## Parallel prefix, generalized

Just as map and reduce are the simplest examples of a common pattern, prefix-sum illustrates a pattern that arises in many, many problems

- Minimum, maximum of all elements to the left of $i$
- Is there an element to the left of $i$ satisfying some property?
- Count of elements to the left of $i$ satisfying some property
- This last one is perfect for an efficient parallel filter ...
- Perfect for building on top of the "parallel prefix trick"


## Parallel Scan

$\operatorname{scan}(\mathrm{o})<x_{0}, \ldots, x_{n-1}>$
==

$$
\left\langle x_{0}, x_{0} \circ x_{1}, \ldots, x_{0} \circ \ldots \circ x_{n-1}\right\rangle
$$

## Operator o

 must be associative!like a fold, except return the folded prefix at each step
pre_scan (o) base <x $x_{0}, \ldots, x_{n-1}>$
==
<base, base o $x_{0}, \ldots$, base o $x_{0} \circ \ldots$ o $x_{n-2}>$
base must be a unit for operator o
sequence with o applied to all items to the left of index in input

## Parallel Filter

Given a sequence input, produce a sequence output containing only elements $v$ such that ( $\mathbf{f} \mathbf{v}$ ) is true

Example: let $\mathrm{f} x=\mathrm{x}>10$

$$
\begin{aligned}
& \text { filter } f<17,4,6,8,11,5,13,19,0,24> \\
= & <17,11,13,19,24>
\end{aligned}
$$

Parallelizable?

- Finding elements for the output is easy
- But getting them in the right place seems hard


## Parallel prefix to the rescue

Use parallel map to compute a bit-vector for true elements:

$$
\begin{aligned}
& \text { input }<17,4,6,8,11,5,13,19,0,24> \\
& \text { bits }<1,0,0,0,1,0,1,1,0,1>
\end{aligned}
$$

Use parallel-prefix sum on the bit-vector:
bitsum <1, 1, 1, 1, 2, 2, 3, 4, 4, 5>

For each i , if bits[i] == 1 then write input[i] to output[bitsum[i]] to produce the final result:

```
output <17, 11, 13, 19, 24>
```

QUICKSORT

## Quicksort review

Recall quicksort was sequential, in-place, expected time $O(n \log n)$

## Best / expected case work

1. Pick a pivot element O(1)
2. Partition all the data into:
$\mathrm{O}(\mathrm{n})$
A. The elements less than the pivot
B. The pivot
C. The elements greater than the pivot
3. Recursively sort A and C
$2 \mathrm{~T}(\mathrm{n} / 2)$

How should we parallelize this?

## Quicksort

Best / expected case work

1. Pick a pivot element O(1)
2. Partition all the data into: $\mathrm{O}(\mathrm{n})$
A. The elements less than the pivot
B. The pivot
C. The elements greater than the pivot
3. Recursively sort A and C
$2 \mathrm{~T}(\mathrm{n} / 2)$

Easy: Do the two recursive calls in parallel

- Work: unchanged. Total: $O(n \log n)$
- Span: now $T(n)=O(n)+1 T(n / 2)=O(n)$


## Doing better

As with mergesort, we get a $O(\log n)$ speed-up with an infinite number of processors. That is a bit underwhelming

- Sort $10^{9}$ elements 30 times faster
(Some) Google searches suggest quicksort cannot do better because the partition cannot be parallelized*
- The Internet has been known to be wrong $)$
- But we need auxiliary storage (no longer in place)
- In practice, constant factors may make it not worth it

Already have everything we need to parallelize the partition...

## Parallel partition (not in place)

Partition all the data into:
A. The elements less than the pivot
B. The pivot
C. The elements greater than the pivot

This is just two filters!

- We know a parallel filter is $O(n)$ work, $O(\log n)$ span
- Parallel filter elements less than pivot into left side of aux array
- Parallel filter elements greater than pivot into right size of aux array
- Put pivot between them and recursively sort

With $O(\log n)$ span for partition, the total best-case and expectedcase span for quicksort is

$$
T(n)=O(\log n)+1 T(n / 2)=O\left(\log ^{2} n\right)
$$

## Example

Step 1: pick pivot as median of three

| 8 | 1 | 4 | 9 | 0 | 3 | 5 | 2 | 7 | 6 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |

Steps 2a and 2c (combinable): filter less than, then filter greater than into a second array


Step 3: Two recursive sorts in parallel

- Can copy back into original array (like in mergesort)


## More Algorithms

- To add multiprecision numbers.
- To evaluate polynomials
- To solve recurrences.
- To implement radix sort
- To delete marked elements from an array
- To dynamically allocate processors
- To perform lexical analysis. For example, to parse a program into tokens.
- To search for regular expressions. For example, to implement the UNIX grep program.
- To implement some tree operations. For example, to find the depth of every vertex in a tree
- To label components in two dimensional images.

See Guy Blelloch "Prefix Sums and Their Applications"

## Summary

- Parallel prefix sums and scans have many applications
- A good algorithm to have in your toolkit!
- Key idea: An algorithm in 2 passes:
- Pass 1: build a "reduce tree" from the bottom up
- Pass 2: compute the prefix top-down, looking at the leftsubchild to help you compute the prefix for the right subchild


## ASSIGNMENT \#7: PROGRAMMING WITH PARALLEL SEQUENCES

## Do the reading . . .

Chapter 2, "Search Engine Indexing"

$\nabla$
$\square$
(Read also Chapter 3, "Page Rank" so you can appreciate what you were doing in Assignment 5...)
(On reserve for this course, available at canvas.princeton.edu, select this course, then "reserves")

## map-reduce API for Assignment 7

|  | Create seq of length $n$, element i holds $f(i)$ | n | 1 |
| :---: | :---: | :---: | :---: |
| seq_of_array: $\alpha$ array -> $\alpha$ seq | Create a sequence from an array | 1 | 1 |
| array_of_seq: $\alpha$ seq -> $\alpha$ array | Create an array from a sequence | 1 | 1 |
| iter (f: $\alpha$-> unit): $\alpha$ seq -> unit | Applying f on each element in order. | n | n |
| length: $\alpha$ seq -> int | Return the length of the sequence | 1 | 1 |
| empty: unit -> $\alpha$ seq | Return the empty sequence | 1 | 1 |
| cons: $\alpha$-> $\alpha$ seq -> $\alpha$ seq | cons a new element on the beginning | n | 1 |
| singleton: $\alpha->\alpha$ seq | Return the sequence with a single element | 1 | 1 |
| append: $\alpha$ seq -> $\alpha$ seq -> $\alpha$ seq | (nondestructively) concatenate two sequences | $m+n$ | 1 |
| nth: $\alpha$ seq -> int -> $\alpha$ | Get the nth value in the sequence. Indexing is zero-based. | 1 | 1 |
| map ( $f: \alpha->\beta$ ) -> $\alpha$ seq -> $\beta$ seq | Map the function $f$ over a sequence | n | 1 |
| $\begin{aligned} & \text { reduce (f: } \alpha->\alpha->\alpha \text { ) (base: } \alpha \text { ): } \\ & \quad \alpha \text { seq }->\alpha \end{aligned}$ | Fold a function $f$ over the sequence. <br> f must be associative, and base must be the unit for f . | n | $\log \mathrm{n}$ |
| $\begin{aligned} & \text { mapreduce: }(\alpha->\beta)->(\beta->\beta->\beta)-> \\ & \beta->\alpha \text { seq }->\beta \end{aligned}$ | Combine the map and reduce functions. | n | $\log n$ |
| flatten: $\alpha$ seq seq -> $\alpha$ seq | flatten [[a0;a1]; [a2;a3]] = [a0;a1;a2;a3] | n | $\log n$ |
| repeat ( $x$ : $\alpha$ ) ( $n$ : int) : $\alpha$ seq | repeat x $4=[x ; x ; x ; x]$ | n | 1 |
| zip: ( $\alpha$ seq ${ }^{*} \beta$ seq) -> ( $\alpha^{*} \beta$ ) seq | zip [a0;a1] [b0;b1;b2] = [(a0,b0);(a1,b1)] | n | 1 |
| split: $\alpha$ seq -> int $->\alpha$ seq * $\alpha$ seq | split [a0;a1;a2;a3] 1= ([a0],[a1;a2;a3]) | n | 1 |
| $\begin{gathered} \text { scan: }(\alpha->\alpha->\alpha)->\alpha-> \\ \alpha \text { seq }->\alpha \text { seq } \end{gathered}$ | ```scan fb [a0;a1;a2;..] = [f b a0; f (f b a0) a1; f (f (f b a0) a1) a2; ...]``` | n | ${ }_{23} \log n$ |

## NESL

## These parallel-sequence operators are inspired by the NESL language (and system) developed by Guy Blelloch. http://www.cs.cmu.edu/~scandal/nesl.html



NESL is a parallel language developed at Carnegie Mellon. It integrates ideas from the theory community (parallel algorithms), the languages community (functional languages) and the systems community (many of the implementation techniques). The most important new ideas behind NESL are

1. Nested data parallelism: this feature offers the benefits of data parallelism, concise code that is easy to understand and debug, while being well suited for irregular algorithms, such as algorithms on trees, graphs or sparse.
2. A language-based performance model: this gives a formal way to calculate the work and depth of a program. These measures can be related to running time on parallel machines.

# IMPLEMENTATION OF PARALLEL SEQUENCES 

## Data Centers: Lots of Connected Computers!



## Real Machines

## Chip



## Real Machines

## Board



## Real Machines



Rack

## Server room



## Real Machines



## s: int seq

length(s) $=10^{9}$

## Real Machines

##  $10^{4}$

$10^{9}$

## Real Machines



## Real Machines



## API for Assignment 7

```
module type S = sig
    type 'a t
    val tabulate : (int -> 'a) -> int -> 'a t
    val seq_of_array : 'a array -> 'a t
    val array_of_seq : 'a t -> 'a array
    val iter: ('a -> unit) -> 'a t -> unit
    val length : 'a t -> int
    val empty : unit ->'a t
    val cons:'a -> 'a t-> 'a t
    val singleton:'a -> 'a t
    val append : 'a t-> 'a t->
    val nth : 'a t -> int -> 'a
    val map : ('a -> 'b) -> 'a t.
    val map_reduce : ('a -> 'b
    val reduce : ('a -> 'a -> 'a)
    val flatten:'a t t-> 'a t
    val repeat : 'a -> int -> 'a
    val zip : ('a t * 'b t) -> ('a
    val split : 'a t -> int -> 'a t
    val scan: ('a -> 'a -> 'a) ->
end
module ArraySeq : S = struct
    type 'a t = 'a array
    let length = Array.length
    let empty () = Array.init 0 (fun _ -> raise (Invalid_argument ""))
    let singleton x = Array.make 1x
    let append = Array.append
    let cons (x:'a) (s:'a t) = append (singleton x) s
    let tabulate f n = Array.init nf
    let nth = Array.get
    let map = Array.map

\section*{Work/Span estimation}
module type S = sig
type 'a t
val tabulate : (int -> 'a) -> int -> ' val seq_of_array : 'a array -> 'a t val array_of_seq : 'a t-> 'a array val iter: ('a -> unit) -> 'a t -> unit val length : 'a t-> int
val empty : unit ->'a t
val cons : 'a -> 'at ->'a t
val singleton : 'a -> 'a t
val append : 'a t-> 'a t-> 'a t
val nth : 'a t-> int -> 'a
val map : ('a -> 'b) -> 'a t-> 'b t
val map_reduce : ('a -> 'b) -> ('b val reduce : ('a -> 'a -> 'a) -> 'a -> val flatten : 'a t t-> 'a t val repeat : 'a -> int -> 'a t val zip : ('a t * 'b t) -> ('a * 'b) t val split : 'a t-> int -> 'a t * 'a t val scan: ('a -> 'a -> 'a) -> 'a -> 'a end
```

module Accounting (M: S) : SCount =
struct

```
    let work = ref 0
    let span = ref 0
    let reporting name \(f x=\)...
    module SM = struct
    type 'a t = 'a M.t
    let tabulate \(\mathrm{f} \mathrm{n}=\) (cost n 1 ;
        let \(s=\) !span in
        let \(\mathrm{smax}=\) ref s in
        let \(\mathrm{z}=\) M.tabulate (fun \(\mathrm{x}->\) let \(\mathrm{y}=\mathrm{f} \mathrm{x}\) in
                        smax := max (!smax) (!span);
span :=s; y) \(n\)
in span := !smax; z)
    let length \(a=\) (cost 11 ; M.length \(a\) )
    let append \(a b=(\operatorname{cost}(M\). length \(a+M\).length \(b) 1\);
                                    M.append a b)
    end
end

\section*{How to use it}
```

Open Sequence module A = Accounting(ArraySeq) module $\mathrm{M}=\mathrm{A} . \mathrm{SM}$
let s1 = M.seq_of_array [|1;2;3;4;5|]
let f (s: int M.seq) = M.map (fun i -> i+1) s
let s2 = A.reporting "test1" f s1
let r = Array.to_list (M.array_of_seq s2)
(* Prints: *)
test1 work=5 span=1
r : int list = [2;3;4;5;6]
let s1 = M.seq_of_array [|1;2;3;4;5|]
let f(s: int M.seq) = M.map (fun i -> i+1) s
let s2 = A.reporting "test1" f s1
let r = Array.to_list (M.array_of_seq s2)
(* Prints: nothing *)
r : int list = [2;3;4;5;6]

```

\section*{Discussion}

How to use these operators to make an inverted index?
key: URL value: contents of page (HTMA)
sequence of words
key: word
value: sequence of (URL, position-in-seq) pairs

\section*{Discussion}

How to use these operators to make an inverted index?
key: URL value: word seq
key: word value: (URL*int) seq

\section*{Discussion}

How to use these operators to make an inverted index?

\section*{key: URL value: word seq \\ (URL * (word seq)) seq}
key: word value: (URL*int) seq

\section*{Discussion}

How to use these operators to make an inverted index?

Input web pages: (URL* (word seq)) seq
key: word value: (URL *int) seq
finite map: word \(\rightarrow\left(\left(\right.\right.\) URL* \({ }^{*}\) int \()\) seq \()\)
Implement by balanced binary search tree (such as 2-3 tree) from OCaml's Map library

\section*{Discussion}

How to use these operators to make an inverted index?

Input web pages: (URL* (word seq)) seq
Now, let's focus on a single web page,
one element of this sequence of web pages

\section*{Discussion}
\[
\begin{array}{lllll}
0 & 1 & 2 & 3 & 4
\end{array}
\]
(URL* (word seq))
(foo.com, [the;play;is;the;thing])

word ((URL*int)seq) Map.t

\section*{Discussion}
(bar.com, [play;the;thing])

play \(\mapsto[(\) bar.com, 0\()]\) the \(\mapsto[(\) bar.com,1)]
thing \(\mapsto[(\) bar.com,2)]
(foo.com, [the;play;is;the;thing])

is \(\mapsto[(\) foo.com, 2\()]\)
play \(\mapsto[(f o o . c o m, 1)]\)
the \(\mapsto[\) (foo.com,0); (foo.com,3)] thing \(\mapsto[(f o o . c o m, 4)]\)

\section*{Discussion}
(bar.com, [play;the;thing])

play \(\mapsto[(\) bar.com, 0\()]\) the \(\mapsto[(\) bar.com,1)] thing \(\mapsto[(\) bar.com,2)]
(foo.com, [the;play;is;the;thing])

is \(\quad \mapsto[(\) foo.com, 2\()]\)
play \(\mapsto[\) (foo.com,1)]
the \(\mapsto[(\) foo.com, 0\() ;\) (foo.com,3)]
thing \(\mapsto[(f o o . c o m, 4)]\)
is \(\quad \mapsto[\) (foo.com,2)]
play \(\mapsto ~[[(b a r . c o m, 0) ; ~(f o o . c o m, 1)]\)
the \(\mapsto[(\) bar.com, 1); (foo.com,0); (foo.com,3)]
thing \(\mapsto ~[[(\) bar.com,2); (foo.com,4)]

\section*{Discussion}

How to use these operators to make an inverted index?

Input web pages: (URL* (word seq)) seq

word ((URL*int)seq) Map.t

\section*{Discussion}

How to use these operators to make an inverted index?

Input web pages: (URL* (word seq)) seq

word ((URL*int)seq) Map.t

This has been a brief introduction to give you a flavor of what you have to do. More details in the homework ... but not necessarily a lot more - you'll have to think for yourself.

And: There is not "one true solution" to this homework.

\section*{Don't "hide" work and span!}

\section*{Open Sequence module A = Accounting(ArraySeq) module \(\mathrm{M}=\mathrm{A} . \mathrm{SM}\)}
let rec costly ( \(n\) : int) \(=\) if \(n=0\) then 1 else costly \((n-1)+\) costly ( \(n-1\) )
let s1 = M.seq_of_array [|51;52;53;54;55|]
let \(\mathrm{f}(\mathrm{s}\) : int M.seq) \(=\mathrm{M} . \operatorname{map}\) costly s
let s2 = A.reporting "test2" f s1
let \(r=\) Array.to_list (M.array_of_seq s2)
(* Prints: *)
test 2 work=5 span=1
\(r\) : int list \(=[2 ; 3 ; 4 ; 5 ; 6]\)

Ideally, each function you write in OCaml should do a small amount of computation (other than nested calls to the M operators).

\section*{CONCLUSION}

\section*{Summary}

By using the Parallel Sequence operators to combine purefunctional implementations of primitive functions, you can:
- Write highly parallel programs
- that scale to many processors
- with fault-tolerance built in
- that compute the same answer deterministically no matter how the parallel execution goes
- while still thinking at a high level of abstraction, independent of the gory details of your parallel machine.```

