

Lecture 2

Parallel Map, Fold, and Scan

Maps

```
def mapSeq[A,B](lst: List[A], f : A => B): List[B] = lst match {  
  case Nil => Nil  
  case h :: t => f(h) :: mapSeq(t,f)  
}
```

```
scala> mapSeq(List(2,3,4), (x:Int) => x*x)  
res1: List[Int] = List(4, 9, 16)
```

List traversal is not parallelizable: we need $O(n)$ span to reach n -th element (even with ∞ threads).

- ▶ makes sense to parallelize only for expensive operation f

Alternatives to Lists

Two simple alternatives:

- ▶ arrays
 - ▶ imperative: must be careful that parallel tasks write to disjoint parts of the array
 - ▶ on a shared memory machine, easy to dynamically partition (just pass around indices for array segments)
 - ▶ ideal memory locality
 - ▶ hard to construct from pieces
- ▶ immutable trees
 - ▶ purely functional, produce new trees
 - ▶ no need to worry about disjointness of writes by parallel tasks
 - ▶ easier to combine
 - ▶ can have high memory allocation overhead
 - ▶ can have bad locality

We start with arrays.

Map on Arrays: Sequential Map on Segment

```
def mapASegSeq[A,B](inp: Array[A], left: Int, right: Int,
                    f : A => B,
                    out: Array[B]) = {
  // Writes to out(i) for left <= i <= right-1
  var i= left
  while (i < right) {
    out(i)= f(inp(i))
    i= i+1
  }
}
val i= Array(2,3,4,5,6)
val o= Array(0,0,0,0,0)
val f=(x:Int) => x*x
mapASegSeq(i, f, 1, 3, o)
o
====>
res1: Array[Int] = Array(0, 9, 16, 0, 0)
```

Parallel Map on Arrays

```
def mapASegPar[A,B](inp: Array[A], left: Int, right: Int,  
    f : A => B,  
    out: Array[B]): Unit = {  
    // Writes to out(i) for left <= i <= right-1
```

Parallel Map on Arrays

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                    out: Array[B]): Unit = {  
  // Writes to out(i) for left <= i <= right-1  
  
  if (right - left < threshold)  
    mapASegSeq(inp, f, left, right, out)  
  else {  
    val mid = left + (right - left)/2  
    val _ = parallel(mapASegPar(inp, left, mid, f, out),  
                    mapASegPar(inp, mid, right, f, out))  
  }  
}
```

Parallel Map on Arrays

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                    mapASegPar(inp, mid, right, f, out))  
  }  
}
```

- ▶ threshold needs to be large enough to compensate expense of parallel task creation
- ▶ we must ensure that two arguments of parallel write to disjoint parts of out array; do they?

Example: Pointwise Exponent of an Array

$$\text{Array}(a_1, a_2, \dots, a_n) \longrightarrow \text{Array}(|a_1|^p, |a_2|^p, \dots, |a_n|^p)$$

We can use previously defined higher-order functions:

```
def power(x: Int, p: Double): Int =  
    math.exp(p * math.log(math.abs(x))/logE).toInt  
def f(x: Int): Double = power(x, p)
```

```
mapASegSeq(inp, 0, inp.length, f, out) // sequential
```

```
mapASegPar(inp, 0, inp.length, f, out) // parallel
```


Exponent of an Array, Inlined Manually - Sequential

```
def mapPowerSeq(inp: Array[Int], p: Double,  
                left: Int, right: Int,  
                out: Array[Double]): Unit = {  
  var i= left  
  while (i < right) {  
    out(i)= power(inp(i),p)  
    i= i+1  
  }  
}
```

Exponent of an Array, Inlined Manually - Parallel

```
def mapPowerPar(inp: Array[Int], p: Double,
               left: Int, right: Int,
               out: Array[Double]): Unit = {
  if (right - left < threshold) {
    var i = left
    while (i < right) {
      out(i) = power(inp(i), p)
      i = i + 1
    }
  } else {
    val mid = left + (right - left) / 2
    val _ = parallel(normsOfPar(inp, p, left, mid, out),
                    normsOfPar(inp, p, mid, right, out))
  }
}
```

Measuring Performance Using Scalometer

- ▶ `inp.length = 2000000`
- ▶ `threshold = 10000`
- ▶ Intel(R) Core(TM) i7-3770K CPU @ 3.50GHz
(4-core, 8 HW threads), 16GB RAM

<i>expression</i>	<i>time(ms)</i>
<i>mapASegSeq(inp, 0, inp.length, f, out)</i>	174.17
<i>mapASegPar(inp, 0, inp.length, f, out)</i>	
<i>mapPowerSeq(inp, p, 0, inp.length, out)</i>	
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- ▶ does parallelization pay off?
- ▶ does manually removing higher-order functions pay off?

Trees

Consider trees where

- ▶ leaves store array segments
- ▶ non-leaf node stores number of elements in left subtree

```
sealed abstract class Tree[A] { val size: Int }  
case class Leaf[A](a: Array[A]) extends Tree[A] {  
  override val size = a.size  
}  
case class Node[A](l: Tree[A], r: Tree[A]) extends Tree[A] {  
  override val size = l.size + r.size  
}
```

Assume our trees are balanced: we can explore branches in parallel

Functional Parallel Map on Trees - Creates New Tree

```
def mapTreePar[A:Manifest,B:Manifest](t: Tree[A], f: A => B) : Tree[B] =  
t match {  
  case Leaf(a) => {  
    val len = a.length  
    val b = new Array[B](len)  
    var i = 0  
    while (i < len) {  
      b(i) = f(a(i))  
      i = i + 1  
    }  
    Leaf(b)  
  }  
  case Node(l,r) => {  
    val (lb,rb) = parallel(mapTreePar(l,f),mapTreePar(r,f))  
    Node(lb, rb)  
  }  
}
```

Speedup and performance similar as the array

Parallel Fold (Reduce) for Associative Operations

Fold / Reduce

If $f(x, y) = x + y$ then

$$\text{List}(2, 5, 20).\text{foldLeft}(100)(f) = ((100 + 2) + 5) + 20$$

Given $f : (B, A) \Rightarrow B$

$$\text{List}(a_1, a_2, a_3).\text{foldLeft}(b)(f) = f(f(f(b, a_1), a_2), a_3)$$

Given $g : (A, B) \Rightarrow B$

$$\text{List}(a_1, a_2, a_3).\text{foldRight}(b)(g) = g(a_1, g(a_2, g(a_3, b)))$$

$$\text{List}(a_1, a_2, a_3).\text{foldRight1}(g) = g(a_1, g(a_2, a_3))$$

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- ▶ iterating arbitrary functions leads to arbitrary messy behavior

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Difficult to parallelize if we know nothing about f :

- ▶ iterating arbitrary functions leads to arbitrary messy behavior

We look at functions $f : (A, A) \Rightarrow A$ that are **associative** and try to fold over a data structure in parallel

Associative Operation

$f : (A, A) \Rightarrow A$ is associative iff for every x, y, z :

$$f(x, f(y, z)) = f(f(x, y), z)$$

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Consequence: consider any two expressions with same list of operands connected with \otimes , but different parentheses. Then these expressions are equal.

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How to prove this precisely?

Lemma about Combining foldRight1 Expressions

Lemma 1: If \otimes is associative, then for non-empty lists xs , ys

$$(xs.foldRight1(\otimes)) \otimes (ys.foldRight1(\otimes)) == (xs ::: ys).foldRight1(\otimes)$$

Proof is by induction on the length of xs .

- ▶ Base case: $xs = List(x)$. Then $xs.foldRight1 = x$ and $xs ::: ys = x :: ys$.

On the right side of equality we have

$(x :: ys).foldRight1(\otimes) = x \otimes (ys.foldRight1(\otimes))$, which is the same as on the left side.

- ▶ Inductive case: $xs = x :: xs1$.

$$\begin{aligned} & ((x :: xs1).foldRight1(\otimes)) \otimes (ys.foldRight1(\otimes)) = \\ & (x \otimes (xs1.foldRight1(\otimes))) \otimes (ys.foldRight1(\otimes)) = \\ & (x \otimes (xs1.foldRight1(\otimes))) \otimes (ys.foldRight1(\otimes)) = (\oplus \text{ assoc}) \\ & x \otimes (xs1.foldRight1(\otimes) \otimes (ys.foldRight1(\otimes))) = (\text{I.H.}) \\ & x \otimes (xs1 ::: ys).foldRight1(\otimes) = \\ & (x :: xs1 ::: ys).foldRight1(\otimes) = (xs ::: ys).foldRight1(\otimes) \end{aligned}$$

Lemma saying every expression equals foldRight1

Lemma 2: If \otimes is associative, then every expression containing only operator \otimes and operands from list xs , in that order, regardless of parantheses, produces the same result for all values as $xs.foldRight1(\otimes)$.

Proof is by induction on the size of the expression, using Lemma 1.

- ▶ Base case: expression has no \oplus so it is of the form x . Then it equals $List(x).foldRight1(\oplus)$
- ▶ Inductive case. Expression is of the form $x \oplus y$. Let xL be the list of operands in x and yL the list of operands in y . By I.H., the expression produces the same result as

$$(xL.foldRight1(\oplus)) \oplus (yL.foldRight1(\oplus))$$

By Lemma 1, this produces the same result as $(xL ::: yL).foldRight1(\oplus)$.

Sequential Fold on Array Segments

```
def foldASegSeq[A,B](inp: Array[A], b0: B,  
                    left: Int, right: Int,  
                    f: (B,A) => B): B = {  
  var b= b0  
  var i= left  
  while (i < right) {  
    b= f(b, inp(i))  
    i= i+1  
  }  
  b  
}
```

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If f was not known to be associative, would this be foldRight or foldLeft ?

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    i= i+1  
  }  
  b  
}
```

If f was not known to be associative, would this be foldRight or foldLeft ?

- ▶ In general, this is foldLeft
- ▶ Result is the same as, e.g., foldRight when f is associative

Parallel Fold on Array Segments

```
def foldASegPar[A](inp: Array[A], a0: A,  
                  left: Int, right: Int,  
                  f: (A,A) => A): A = {  
  // requires f to be associative  
  if (right - left < threshold)  
    foldASegSeq(inp, a0, left, right, f)  
  else {  
    val mid = left + (right - left)/2  
    val (a1,a2) = parallel(foldASegPar(inp, a0, left, mid, f),  
                          foldASegPar(inp, a0, mid, right, f))  
    f(a1,a2)  
  }  
}
```

- ▶ Sequential version computed *foldLeft*
- ▶ here we compute a more balanced expression tree, combining fold of two halves of array
- ▶ the result is the same when *f* is associative.

Important Associative Operations that Happen to be also Commutative

Examples

- ▶ addition and multiplication modulo a positive integer (e.g. 2^{32}), including the usual arithmetic on 32-bit or 64-bit integers
- ▶ addition and multiplication of BigInt-s (mathematical integers)
- ▶ union, intersection, and symmetric difference of sets
- ▶ boolean operations $\&\&$, $\|\|$, exclusive or

If operation $f(x, y)$ is associative, then \bar{f} defined by

$$\bar{f}((x_1, \dots, x_n), (y_1, \dots, y_n)) = (f(x_1, y_1), \dots, f(x_n, y_n))$$

is also associative.

Associativity Does Not Imply Commutativity

Associativity **does NOT imply** $x \otimes y = y \otimes x$ (commutativity)

Examples of *associative* and *not commutative* operations:

- ▶ concatenation (append) of lists $(x ::: y) ::: z == x ::: (y ::: z)$
and strings
- ▶ matrix multiplication AB
- ▶ composition of relations
 $r \odot s = \{(a, c) \mid \exists b. (a, b) \in r \wedge (b, c) \in s\}$
- ▶ composition of functions $(f \circ g)(x) = f(g(x))$

Similarly, Commutativity Does Not Imply Associativity

Example:

$$f(x, y) = x^2 + y^2 = f(y, x)$$

Then

$$f(f(x, y), z) = (x^2 + y^2)^2 + z^2$$

whereas

$$f(x, f(y, z)) = x^2 + (y^2 + z^2)^2$$

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In general, if $p(x, y)$ is commutative and $h_1(z), h_2(z)$ are arbitrary, then any function defined by

$$q(x, y) = h_2(p(h_1(x), h_1(y)))$$

is equal to $h_2(p(h_1(y), h_1(x))) = q(y, x)$, so it is commutative, but can lose associativity even if q was associative.

Floating Point Addition is Not Associative

```
scala> val e = 1e-200  
e: Double = 1.0E-200
```

```
scala> val x = 1e200  
x: Double = 1.0E200
```

```
scala> val mx = -x  
mx: Double = -1.0E200
```

```
scala> (x + mx) + e  
res2: Double = 1.0E-200
```

```
scala> x + (mx + e)  
res3: Double = 0.0
```

```
scala> (x + mx) + e == x + (mx + e)  
res4: Boolean = false
```

Similarly for multiplication

Two Rules Implying Associativity

Suppose that $f(x, y)$ is commutative and if we define

$$E(x, y, z) = f(f(x, y), z)$$

then $E(x, y, z) = E(y, z, x)$. Show that f is then associative.

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Solution:

Two Rules Implying Associativity

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Solution:

$$f(f(x, y), z) = f(f(y, z), x) = f(x, f(y, z))$$

Is this Operation on Real Numbers Associative?

Let u, v range over the open interval of reals $(-1, 1)$

$$f(u, v) = \frac{u + v}{1 + uv}$$

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Next

$$f(f(u, v), w) = \frac{\frac{u+v}{1+uv} + w}{1 + \frac{u+v}{1+uv}w} = \frac{u + v + w + uvw}{1 + uv + uw + vw}$$

From the above two we have with v, w, u playing the role of u, v, w :

$$f(u, f(v, w)) = f(f(v, w), u) = \frac{v + w + u + vwu}{1 + vw + vu + uv}$$

So two sides of associativity condition are equal.

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What is the motivation for this operation?

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So two sides of associativity condition are equal.

What is the motivation for this operation?

Law of adding (normalized) velocities in special relativity theory

Velocity Addition as an Example for Fold

```
val c = 2.99792458e8
def assocOp(v1: Double, v2: Double): Double = {
  val u1 = v1/c
  val u2 = v2/c
  (v1 + v2)/(1 + u1*u2)
}
def addVelSeq(inp: Array[Double]): Double = {
  foldASegSeq(inp, 0.0, 0, inp.length, assocOp)
}
def addVelPar(inp: Array[Double]): Double = {
  foldASegPar(inp, 0.0, 0, inp.length, assocOp)
}
```

We obtain noticeable speedup (2-3 times).

Value computed differs slightly because of roundoff errors.

A Family of Associative Operations on Sets

Define binary operation on sets A, B by

$$f(A, B) = (A \cup B)^*$$

where $*$ is any operator on sets (closure) with these properties:

- ▶ $A \subseteq A^*$
- ▶ if $A \subseteq B$ then $A^* \subseteq B^*$
- ▶ $(A^*)^* = A^*$

Prove that f is associative.

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Hint:

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where $*$ is any operator on sets (closure) with these properties:

- ▶ $A \subseteq A^*$
- ▶ if $A \subseteq B$ then $A^* \subseteq B^*$
- ▶ $(A^*)^* = A^*$

Prove that f is associative.

Hint:

- ▶ Observe that $P \subseteq Q^*$ implies $P^* \subseteq Q^*$

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- ▶ Use previous to show $f(f(A, B), C) = (A \cup B \cup C)^*$
- ▶ Observe that $f(f(A, B), C) = f(f(B, C), A)$ and $f(A, B) = f(B, A)$, then use the slide “two rules implying associativity”.

Parallel Scan

Scan

If $f(x, y) = x + y$ then

$$\text{List}(2, 5, 20).\text{scanLeft}(100)(f) = \text{List}(100, 102, 107, 127)$$

$$\text{List}(a_1, a_2, a_3).\text{scanLeft}(f)(a_0) = \text{List}(b_0, b_1, b_2, b_3)$$

where

- ▶ $b_0 = a_0$
- ▶ $b_1 = f(b_0, a_1)$
- ▶ $b_2 = f(b_1, a_2)$
- ▶ $b_3 = f(b_2, a_3)$

Can scan be parallelized? Assume f associative.

Sequential Scan over a Segment

```
def scanASegSeq1[A](inp: Array[A], left: Int, right: Int,
                    a0: A, f: (A,A) => A,
                    out: Array[A]) = {
  if (left < right) {
    var i = left
    var a = a0
    while (i < right) {
      a = f(a, inp(i))
      out(i+1) = a
      i = i+1
    }
  }
}
```

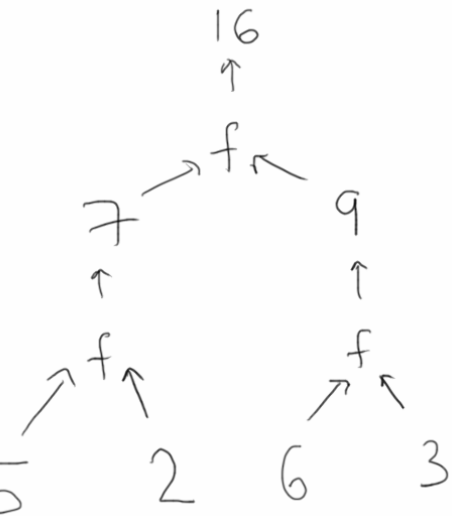
Scans array segment $\text{inp}(\text{left})$ to $\text{inp}(\text{right}-1)$, storing results into $\text{out}(\text{left}+1)$ to $\text{out}(\text{right})$. At the end, each $\text{out}(i+1)$ stores fold of elements: $[\text{a0}, \text{in}(\text{left}), \dots, \text{in}(i)]$ for i from left to $\text{right}-1$. In particular, $\text{out}(\text{left}+1)$ stores $f(\text{a0}, \text{inp}(\text{left}))$ and $\text{out}(\text{right})$ stores fold of $[\text{a0}, \text{in}(\text{left}), \dots, \text{inp}(\text{right}-1)]$. The value a0 is not directly stored anywhere.

Computation Tree for Recording Results of Parallel Fold

```
sealed abstract class FoldTree[A] {  
  val res: A // value of the tree, whether it is leaf or not  
}  
case class Leaf[A](from: Int, to: Int, resLeaf: A) extends FoldTree[A] {  
  val res= resLeaf  
}  
case class Node[A](l: FoldTree[A], r: FoldTree[A],  
                  resNode: A) extends FoldTree[A] {  
  val res= resNode  
}
```


upsweep: Parallel Fold that Records its Computation Tree

```
def upsweep[A](inp: Array[A], left: Int, right: Int,
               a0: A, f: (A,A) => A): FoldTree[A] = {
  // requires f to be associative
  if (right - left < threshold)
    Leaf(left, right, foldASegSeq(inp, left, right, a0, f))
  else {
    val mid = left + (right - left)/2
    val (t1,t2) = parallel(upsweep(inp, left, mid, a0, f),
                          upsweep(inp, mid, right, a0, f))
    Node(t1, t2, f(t1.res,t2.res))
  } }
```

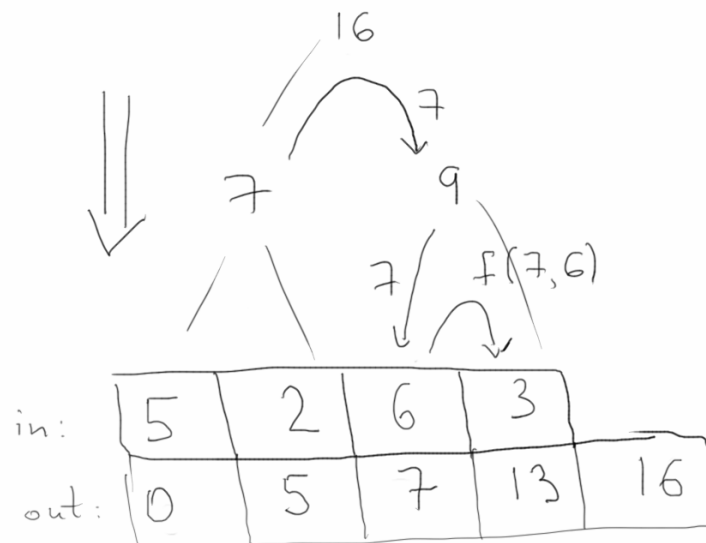


Folds array segment in parallel and record the intermediate computation results in a `Tree[A]`.

- ▶ In the context of scan, this phase is called **upsweep**. For an intuition, picture the array to fold on the bottom, and the root of the tree at the top. Once the 'parallel' tasks are initiated, the results are combined in the 'up' direction, from array to the result of the fold.

Using the Tree to Compute Scan: Key Part

```
def downsweep[A](inp: Array[A],
                  a0: A, f: (A,A) => A,
                  t: FoldTree[A],
                  out: Array[A]): Unit = {
  t match {
  case Leaf(from, to, res) =>
    scanASegSeq1(inp, from, to, a0, f, out)
  case Node(l, r, res) => {
    parallel(downsweep(inp, a0, f, l, out),
             downsweep(inp, f(a0, l.res), f, r, out))
  }
  }
}
```



Scan of [5,2,6,3] with +

Scanning the Entire Array

```
def scanASegPar[A](inp: Array[A], from: Int, to: Int,  
                  a0: A, f: (A,A) => A, out: Array[A]) = {  
  val t = upsweep(inp, from, to, a0, f)  
  downsweep(inp, a0, f, t, out) }  

```

```
def scanAPar[A](inp: Array[A], a0: A, f: (A,A) => A,  
               out: Array[A]) = {  
  out(0) = a0  
  scanASegPar(inp, 0, inp.length, a0, f, out) }  

```

Example: producing all partial sums of velocities, relativistically:

```
val c = 2.99792458e8  
def assocOp(v1: Double, v2: Double): Double = {  
  val u1 = v1/c; val u2 = v2/c  
  (v1 + v2)/(1 + u1*u2)  
}  
scanAPar(inp, 0.0, assocOp, outPar)
```

Combining and Fusing Operations

Combining Maps

If $f(x, y) = x + y$ then

$$List(2, 5, 20).map(f).map(g) == List(g(f(2)), g(f(5)), g(f(20)))$$

Instead of producing intermediate structure, we can apply both f and g in one pass.

- ▶ this idea applies to both sequential and parallel traversals

Array Norm

$$\sum_{i=s}^{t-1} [|a_i|^p]$$

Which combination of operations does sum of powers correspond to?

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- ▶ second: $fold$ with $+$

Note: folding with $f(x, y) = |x|^p + |y|^p$ gives

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- ▶ first: $map(x \Rightarrow power(abs(x), p))$
- ▶ second: $fold$ with $+$

Note: folding with $f(x, y) = |x|^p + |y|^p$ gives nonsense

Array Norm as Example of Fused Operations

The recursive case is not affected, as if we had just fold of $+$:

```
def pNormRec(a: Array[Int], p: Real): Int =  
  power(segmentRec(a, p, 0, a.length), 1/p)
```

```
def segmentRec(a: Array[Int], p: Real, s: Int, t: Int) = {  
  if (t - s < threshold)  
    sumSegment(xs, p, s, t) // we read array content only here  
  else {  
    val mid = s + (t - s)/2  
    val (leftSum, rightSum) =  
      parallel(segmentRec(a, p, s, mid),  
              segmentRec(a, p, mid, t))  
    leftSum + rightSum  
  }  
}
```

Array Norm as Example of Fused Operations

fused map and sum happen below the cutoff:

```
def sumSegment(a: Array[Int], p: Double, s: Int, t: Int): Int = {  
  var i = s; var sum: Int = 0  
  while (i < t) {  
    sum = sum + power(a(i), p) // fused map(power(-,p)) and fold(-+)  
    i = i + 1  
  }  
  sum  
}
```

Histograms

Suppose elements of $inp : Array[Int]$ are between 0 and 99.

For each interval $[10k, 10k + 9]$, count how many array elements are in there, storing the result in array $hist : Array$ of size 10.

Express this task as a combination of map and fold.

What is the type on which fold works?

Running Average

Given an array inp , compute array out where $out(i)$ is the average of elements $inp(j)$ for $0 \leq j < i$.

Running Average

Given an array *inp*, compute array *out* where *out*(*i*) is the average of elements *inp*(*j*) for $0 \leq j < i$.

// code from parallel scan

```
def scanASegSeq1[A](inp: Array[A],
                    left: Int, right: Int,
                    a0: A,
                    f: (A,A) => A,
                    out: Array[A]) = {
  if (left < right) {
    var i= left
    var a= a0
    while (i < right) {
      a= f(a,inp(i))
      out(i+1)=a // can we modify it to compute average?
      i= i+1
    }
  }
}
```