Repa Regular, Shape-polymorphic, Parallel Arrays

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Example Applications

Solving the Laplace Equation



Fast Fourier Transform (highpass filter)



Example Applications

Solving the Laplace Equation





Laplace Equation

boundary conditions



5000 steps





2D Fast Fourier Transform (FFT)



Regular, Shape-polymorphic, Parallel Arrays

• Regular Arrays

Arrays are dense and rectangular. Most elements are non-zero.

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Shape Polymorphic

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• Flat Data Parallelism

Individual computations don't need to communicate. Parallel computations don't spark further parallel computations transpose2D

:: Elt e => Array DIM2 e -> Array DIM2 e

transpose2D arr
= backpermute newExtent swap arr
where swap (Z :.i :.j) = Z :.j :.i
 newExtent = swap (extent arr)

10	20	30	
44	55	66	

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- The ordering of the elements changes, but the values do not.

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- An Index Space Transform
- The ordering of the elements changes, but the values do not.
- We usually want to push such transforms into the consumer.

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Matrix Multiplication $(A.B)_{i,j} = \sum_{k=1}^{n} A_{i,k} \cdot B_{k,j}$

a 11	a ₁₂	a ₁₃			1	l	c_{11}	C 12
a 21	an	an		b ₁₁	b ₁₂		a	a
	u22	u23	•	b ₂₁	b 22	=	C ₂₁	C ₂₂
a ₃₁	a ₃₂	a ₃₃		b 21	haa		C ₃₁	C ₃₂
a ₄₁	a ₄₂	a 43		~31	~ 32		C ₄₁	C ₄₂

Matrix Multiplication $(A.B)_{i,j} = \sum_{k=1}^{n} A_{i,k} \cdot B_{k,j}$



• All elements of the result can be computed in parallel!



- trr = transpose2D brr
- arrR = replicate (Z :.All :.colsB :.All) arr brrR = replicate (Z :.rowsA :.All :.All) trr

🗒 c11

c21

- (Z :. colsA :. rowsA) = extent arr
- (Z :. colsB :. rowsB) = extent brr

It's nice to program with bulk operations
 ... but we usually want them to be fused.

• We imagine replicating the source arrays being replicated when writing the program, but we don't want this at runtime.

• Fusion eliminates the intermediate arrays and the corresponding memory traffic.

- Manifest wraps a bona-fide unboxed array. Bulk-strict semantics. Forcing one element forces them all.
- **Delayed** wraps an element producing function, perhaps an index transformation that references some other array.
- Delayed functions are inlined and fused by the existing GHC optimiser (and lots of rewrite rules).

let arr = ...
 brr = map f arr
in mmMult brr brr

data Array sh e
= Manifest sh (UArr e)
| Delayed sh (sh -> e)

let arr = ... brr = map f arr in mmMult brr brr force :: Array sh e -> Array sh e

```
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```

let arr = ...

brr = **force** (map f arr)

- in mmMult brr brr
- For Manifest arrays, force is the identity.
- For Delayed arrays, it evaluates all the elements in parallel, producing a manifest array.
- The programmer must add force manually.

Using the force ...

a 11 a	1 12	a 13					C 11	C 12
		2		b ₁₁	b ₁₂			
	422	a 23	•	b ₂₁	b 22	=	C ₂₁	C ₂₂
a ₃₁ a	1 32	a 33		b 21	b 22		C 31	C 32
a ₄₁ a	112	a ₄₂		~31	₩32		C ₄₁	C ₄₂

 We get better cache performance when accessing the **b** elements left-to-right rather than top-to-bottom



mmMult

- :: (Num e, Elt e)
- => Array DIM2 e
- -> Array DIM2 e -> Array DIM2 e



Replicate and Slice are duals.





- Replicate and Slice are index transforms.
- The values of the array elements do not change.

Type hackery



F

Type hackery











Type hackery





E

F





slice :: (Slice sl, Elt e , Shape (FullShape sl)) , Shape (SliceShape sl)) => Array (FullShape sl) e -> sl -> Array (SliceShape sl) e

Other operations

map	<pre>:: (Shape sh, Elt a, Elt b) => (a -> b) -> Array sh a -> Array sh b</pre>
zip	<pre>:: (Shape sh, Elt a, Elt b) => Array sh a -> Array sh b -> Array sh (a, b)</pre>
foldl	<pre>:: (Shape sh, Elt a, Elt b) => (a -> b -> a)</pre>

reshape :: (Shape sh, Shape sh', Elt e)
=> sh -> Array sh' e -> Array sh e

Matrix Multiplication 1024x1024

on a 2x Quad-core 3Ghz Xenon speedup threads

	GCC	single thread	fastest parallel
Xenon	3.8s	4.6s	0.64s
T2	52s	92s	2.4s

on a 1.4Ghz UltraSPARC T2



- C reference version uses double nested loops.
- Exposing sufficient parallelism on the T2 is a must.

Laplace Equation



	GCC	single thread	fastest parallel
Xenon	0.70	1.7s	0.68s
T 2	6.5s	32s	3.8s

on a 1.4Ghz UltraSPARC T2



- GHC native code generator does no instruction reordering on SPARC. No LLVM 'port.
- Single threaded on T2 is slow

2D Fast Fourier Transform (FFT)

speedup Ô threads

on 2x Quad-core 3GHz Xenon



	GCC	single thread	fastest parallel
Xenon	0.24	8.8s	2.0s
T2	2.4s	98s	7.7s

on 1.4Ghz UltraSPARC T2



- C version is FFTW which uses in-place deep magic.
- Parallelism is no substitute for a better algorithm.

2D Fast Fourier Transform (FFT)



fft1D :: Array (sh:.Int) Double
 -> Array (sh:.Int) Double
 -> Array (sh:.Int) Double

fft1D rofu
| n > 2 = (left +^ right) :+: (left -^ right)
| n == 2 = traverse v id swivel
where
(_ :. n) = extent v
swivel f (ix:.0) = f (ix:.0) + f (ix:.1)
swivel f (ix:.1) = f (ix:.0) - f (ix:.1)
rofu' = evenHalf rofu
left = force . . .fft1D rofu' .evenHalf \$ v

right = force .(*^ rofu).fft1D rofu' .oddHalf \$ v



• The examples we've presented are easy to write, but are cache naive.

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- Using, block based matrix multiplication imposes a restriction on the order of evaluation...
 ... and makes it less obvious how to parallelise.

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- Using, block based matrix multiplication imposes a restriction on the order of evaluation...
 ... and makes it less obvious how to parallelise.
- Repeated index computations can be expensive.
 GHC does not perform strength reduction on its loops.

http://trac.haskell.org/repa

- We depend on the the current version of the GHC head for decent performance.
- There will be a new release in a few weeks with GHC 7.0
- Send me your programs and I'll add them to our performance regression testsuite!

Questions?

Spare Slides

Shapes and Indices

```
data Z
           = Z
data tail :. head = tail :. head
type DIMO = Z
type DIM1 = DIM0 :. Int
type DIM2 = DIM1 :. Int
. . .
class Shape sh where
 rank :: sh -> Int
 size :: sh -> Int
 toIndex :: sh -> sh -> Int
 fromIndex :: sh -> Int -> sh
```

instance Shape Shape Z where ...
instance Shape sh => Shape (sh :. Int) where ...

Generic Traversal and the 3rd order function

```
traverse
:: (Shape sh, Shape sh', Elt e)
=> Array sh e
-> (sh -> sh') -- shape transform
-> ((sh -> e) -> sh' -> e') -- elem transform
-> Array sh' e'
```

- A sane use for a third order function!
- Traverse takes a function to calculate the elements of the array.
- That function is passed a function to get elements of the source array.



```
stencil :: Array DIM2 Double
   -> Array DIM2 Double
```

```
stencil arr
= traverse arr id update
where
  :. height :. width = extent arr
  update get d@(sh :. i :. j)
   = if isBoundary i j
      then get d
      else (get (sh :. (i-1)) :. j)
         + get (sh :. i :. (j-1))
         + get (sh :. (i+1)) :. j)
         + get (sh :. i :. (j+1))) / 4
```