GPGPU Programming in Haskell with Accelerate

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https://github.com/AccelerateHS

Context

- YOW LambdaJam is coming up: May 16-17, Brisbane
- buuut... I need to prepare a talk for it:
 - Split into 30 min + 1.5 hr session
 - Looking for feedback on topics to include / focus on / skip / etc.
 - Particularly if you have tried to use Accelerate before
 - Especially if you have tried and failed

What is GPGPU Programming?

- General Purpose Programming on Graphics Processing Units (GPUs)
- Using your graphics card for something other than playing games
- GPUs have many more cores than a CPU
 - GeForce GTX Titan
 - 2688 cores @ 837 MHz
 - 6 GB memory @ 288 GB/s



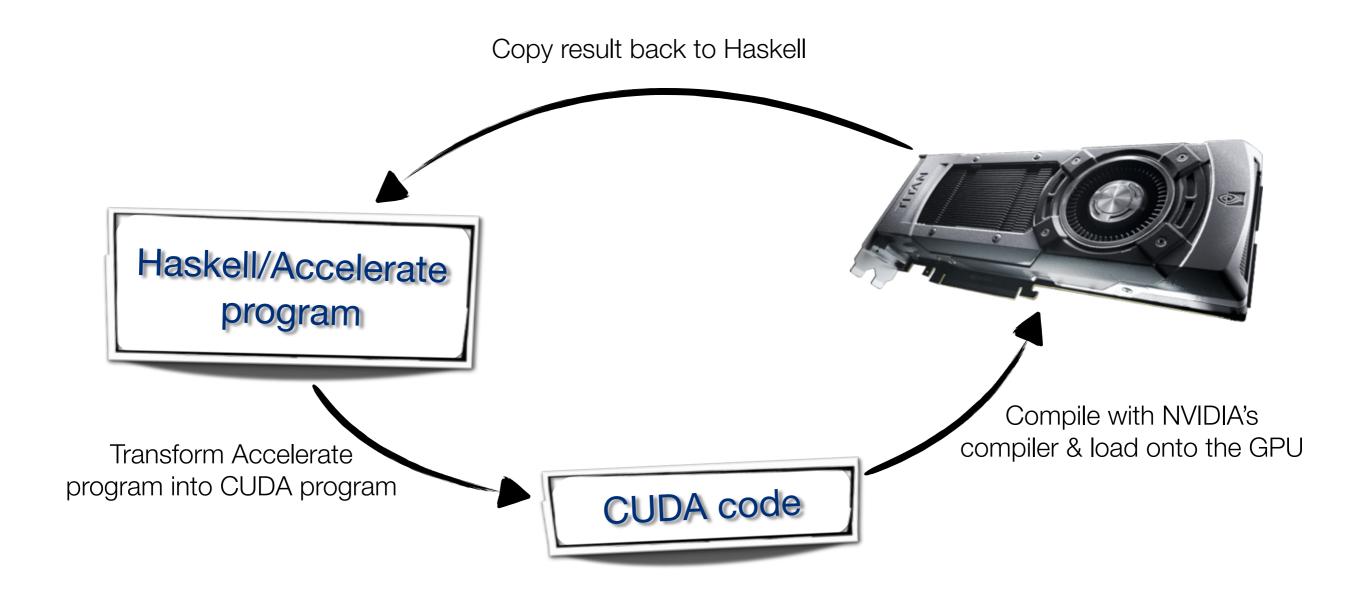
What is GPGPU Programming?

Main differences:

- Single program multiple data (SPMD / SIMD), or just data-parallelism
- All the cores run the same program, but on different data
- We can't program these in the same way as a CPU
 - Different instruction sets: can't run a Haskell program directly
 - More restrictive hardware designs, limited control structures
- GPUs have their own memory
 - Data has to be explicitly moved back and forth

Accelerate

Accelerate is a Domain-Specific Language for GPU programming



Accelerate

- Accelerate is a Domain-Specific Language for GPU programming
 - This process may happen several times during program execution
 - Code and data fragments get cached and reused
- An Accelerate program is a Haskell program the generates a CUDA program
 - However, in many respects this still looks like a Haskell program
 - Shares various concepts with Repa, a Haskell array library for CPUs

Accelerate

- Accelerate computations take place on arrays
 - Parallelism is introduced in the form of collective operations over arrays



- Arrays have two type parameters
 - The shape of the array, or dimensionality
 - The element type of the array: Int, Float, etc.

data Array sh e

Shapes

- Shapes determine the dimensions of the array and the type of the index
 - Z represents a rank-zero array (singleton array with one element)
 - (:.) increases the rank by adding a new dimension on the right
- Examples:
 - One-dimensional array (vector) indexed by Int: (Z :. Int)
 - Two-dimensional array, indexed by Int: (Z :. Int :. Int)
- This style is used at both the type and value level:

```
sh :: Z :. Int
sh = Z :. 10
```

Shapes

We have some handy synonyms as well:

```
type DIM0 = Z
type DIM1 = DIM0 :. Int
type DIM2 = DIM1 :. Int
type DIM3 = DIM2 :. Int
-- and so on...

type Scalar e = Array DIM0 e
type Vector e = Array DIM1 e
```

- Supported array element types are members of the E1t class:
 - ()
 - Int, Int32, Int64, Word, Word32, Word64...
 - Float, Double
 - Char
 - Bool
 - Tuples up to 9-tuples of these, including nested tuples
- Note that Array itself is not an allowable element type. There are no nested arrays in Accelerate, regular arrays only!

Create an array from a list:

```
fromList :: (Shape sh, Elt e) => sh -> [e] -> Array sh e
```

- Generates a multidimensional array by consuming elements from the list and adding them to the array in row-major order

Example:

```
ghci> fromList (Z:.10) [1..10]
```

- Defaulting does not apply, because Shape is not a standard class

```
<interactive>:3:1:
   No instance for (Shape (Z :. head0))
     arising from a use of `fromList'
   The type variable `head0' is ambiguous
   Possible fix: add a type signature that fixes these type
   Note: there is a potential instance available:
     instance Shape sh => Shape (sh :. Int)
       -- Defined in `Data.Array.Accelerate.Array.Sugar'
   Possible fix: add an instance deal
                                                       e (Z
                Number 1 tip:
                                                       .. 10
           Add type signatures
                                        rom the literal `10
   The type variable `head0' is ambiguous
   Possible fix: add a type signature that fixes these type
   Note: there are several potential instances:
     instance Num Double -- Defined in `GHC.Float'
     instance Num Float -- Defined in `GHC.Float'
     instance Integral a => Num (GHC.Real.Ratio a)
       -- Defined in `GHC.Real'
      ...plus 12 others
   In the second argument of `(:.)', namely `10'
```

Create an array from a list:

```
> fromList (Z:.10) [1..10] :: Vector Float
Array (Z :. 10) [1.0,2.0,3.0,4.0,5.0,6.0,7.0,8.0,9.0,10.0]
```

- Multidimensional arrays are similar:
 - Elements are filled along the right-most dimension first

```
> fromList (Z:.3:.5) [1..] :: Array DIM2 Int
Array (Z :. 3 :. 5) [1,2,3,4,5,6,7,8,9,10,11,12,13,14,15]
```

I	2	3	4	5
6	7	8	9	10
П	12	13	14	15

Array indices start counting from zero

```
> let mat = fromList (Z:.3:.5) [1..] :: Array DIM2 Int
> indexArray mat (Z:.2:.1)
12
```

ı	2	3	4	5
6	7	8	9	10
П	12	13	14	15

- Similarly, an array of (possibly nested) tuples:
 - This is just a trick: internally converted into a tuple of arrays

```
> fromList (Z:.2:.3) $ P.zip [1..] ['a'..] :: Array DIM2 (Int,Char)
Array (Z :. 2 :. 3) [(1,'a'),(2,'b'),(3,'c'),(4,'d'),(5,'e'),(6,'f')]
```

	Ι	2	3		a	ь	с	\
•	4	5	6	,	d	e	f	

Data.Array.Accelerate

- Need to import both the base library as well as a backend
 - There is also an interpreter available for testing
 - Runs without using the GPU (much more slowly of course)

```
import Prelude
import Data.Array.Accelerate
import Data.Array.Accelerate.CUDA as CUDA
```

Data.Array.Accelerate

To actually run an Accelerate computation:

```
run :: Arrays a => Acc a -> a
```

- Run comes from whichever backend we have chosen (CUDA)
- Arrays constrains the result to be an Array, or tuple thereof
- What is Acc?
 - This is our DSL type
 - A data structure representing a computation that once executed will yield a result of type 'a'

Accelerate is a library of collective operations over arrays of type Acc

Data.Array.Accelerate

To get arrays into the Acc world:

```
use :: Arrays arrays => arrays -> Acc arrays
```

- This may involve copying data to the GPU
- use injects arrays into our DSL
- run executes the computation to get arrays out
- Using Accelerate focuses on everything in between

Collective Operations

Example: add one to each element of an array

```
> let arr = fromList (Z:.3:.5) [1..] :: Array DIM2 Int
> run $ A.map (+1) (use arr)
Array (Z :. 3 :. 5) [2,3,4,5,6,7,8,9,10,11,12,13,14,15,16]
```

What is the type of map?

Supported shape & element types

```
map :: (Shape sh, Elt a, Elt b)
=> (Exp a -> Exp b)
-> Acc (Array sh a)
-> Acc (Array sh b)
```

Function to apply at every element. But what is Exp?

DSL array

A Stratified Language

- Accelerate is split into two worlds: Acc and Exp
 - Acc represents collective operations over instances of Arrays
 - Exp is a scalar computation on things of type Elt
- Collective operations in Acc comprise many scalar operations in Exp, executed in parallel over Arrays
- Scalar operations can not contain collective operations
 - This excludes nested data parallelism

Scalar Expressions

The type class overloading trick is used for standard Haskell classes

```
(+1) :: (Elt a, IsNum a) => Exp a -> Exp a
```

- Standard boolean operations are available with slightly different names
 - The standard names can not be overloaded

```
(==*) :: (Elt t, IsScalar t) => Exp t -> Exp t -> Exp Bool (/=*), (<*), (>*), min, max, (||*), (&&*) -- and so on...
```

- Conditionals
 - Use sparingly: leads to SIMD divergence

```
(?) :: Elt t => Exp Bool -> (Exp t, Exp t) -> Exp t
```

Scalar Expressions

Bring a Haskell value into Exp land

```
constant :: Elt e -> e -> Exp e
```

Lift an expression into a singleton array

```
unit :: Exp e -> Acc (Scalar e)
```

Reductions

- Folding (+) over a vector produces a sum
 - The result is a one-element array (scalar). Why?

```
> let xs = fromList (Z:.10) [1..] :: Vector Int
> run $ A.fold (+) 0 (use xs)
Array (Z) [55]
```

Fold has an interesting type:

```
fold :: (Shape sh, Elt a)
=> (Exp a -> Exp a -> Exp a)
-> Exp a
-> Acc (Array (sh:.Int) a)
-> Acc (Array sh a)
```

outer dimension removed

Reductions

Fold occurs over the outer dimension of the array

```
> let mat = fromList (Z:.3:.5) [1..] :: Array DIM2 Int
> run $ A.fold (+) 0 (use mat)
Array (Z :. 3) [15,40,65]
```

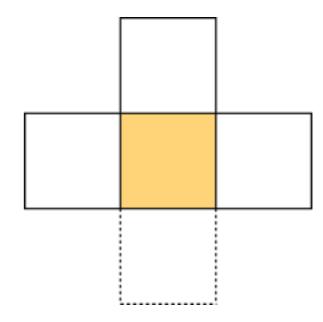
I	2	3	4	5	15
6	7	8	9	10	40
11	12	13	14	15	65

Reductions

- Is this a left-fold or a right-fold?
 - Neither! The fold happens in parallel, tree-like
 - Therefore the function must be associative: (Exp a -> Exp a -> Exp a)
 - (We pretend that floating point operations are associative, though strictly speaking they are not)

Stencils

• A 2D convolution...



Permutations

• Forward / backward permutations ...

• Replicate / slice ...

Problem Solving with Accelerate

Example: password "recovery"

MD5 Algorithm

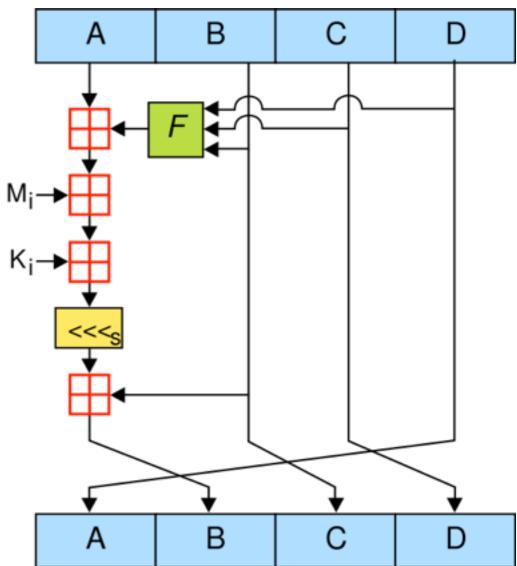
• Aim:

- Implement one round of MD5: unsalted, single 512-bit block
- Apply to an array of words
- Compare hashes to some unknown hash
- i.e. standard dictionary attack
- For fun: send me your (low quality) password hashes
 - tmcdonell@cse.unsw.edu.au

\$ md5 -q -s password
5f4dcc3b5aa765d61d8327deb882cf99

MD5 Algorithm

- Algorithm operates on a 4-word state,
 A, B, C, and D
- There are 4 x 16 rounds: F, G, H, and I
 - M_i is a word from the input message
 - K_i is a constant
 - <<<s is left rotate, by some constant ri
- Each round operates on the 512-bit message block, modifying the state



- Accelerate is a meta programming language
 - Use regular Haskell to generate the expression for each step of the round
 - Produces an unrolled loop

- The constants k_i and r_i can be embedded directly
 - The simple list lookup would be death in standard Haskell
 - Generating the expression need not be performant, only executing it

```
k :: Int -> Exp Word32
k i = constant (ks P.!! i)
  where
    ks = [ 0xd76aa478, 0xe8c7b756, 0x242070db, 0xc1bdceee
    , ...
```

- The message M is stored as an array, so we need array indexing
 - Be wary, arbitrary array indexing can kill performance...

```
(!) :: (Shape ix, Elt e) => Acc (Array ix e) -> Exp ix -> Exp e
```

Get the right word of the message for the given round

```
m :: Int -> Exp Word32
m i
    | i < 16 = msg A.! index1 (constant i)
    | i < 32 = msg A.! index1 (constant ((5*i + 1) `rem` 16))
    | ...</pre>
```

Finally, the non-linear functions F, G, H, and I

```
round :: ABDC -> Int -> ABCD
round (a,b,c,d) i
  | i < 16 = shfl (f b c d)
  | ...
  where
    shfl x = (d, b + ((a + x + k i + m i) `rotateL` r i), b, c)
    f x y z = (x .&. y) .|. ((complement x) .&. z)
    ...</pre>
```

- MD5 applied to a single 16-word vector: no parallelism here
- Lift this operation to an array of n words: stored as a (Z:.16:.n) array
 - Store one word per row column, process many words in parallel
 - Need to use generate, the most general form of array construction. Equivalently, the most easily misused (as we will won't see)

```
generate :: (Shape sh, Elt e)
    => Exp sh
    -> (Exp sh -> Elt e)
    -> Acc (Array sh e)
```

- As always, data layout is important
 - Accelerate arrays are stored in row-major order
- CUDA threads work together
 - generate uses one thread per element
 - For best performance threads need to index adjacent elements of an array
 - This only works for m_i if all the first letters are adjacent in memory, etc.

STOP: demo time

Problem Solving Accelerate

Hints for when things don't work as you expect

Executing Computations

• run vs. run1

Acc Inside

- · We can see the data structure Accelerate generates by omitting 'run'
 - Useful to check if intermediates have been fused away

```
> let xs = fromList (Z:.10) [1..] :: Vector Int
> A.map (*2) $ A.map (\x -> x `mod` 2 ==* 0 ? (x-2, x+3)) (use xs)
let a0 = use (Array (Z :. 10) [1,2,3,4,5,6,7,8,9,10])
in
map (\x0 -> 2 * (0 ==* (mod (x0, 2)) ? (-2 + x0, 3 + x0))) a0
```

Embedded Scalars

• Check what & how often code is getting compiled: -ddump-cc

Nested Data-Parallelism

• matrix-vector multiply: using replicate & a higher-dimensional fold

Iterations

- Running out of memory: Floyd-Warshall
- Using the (>->) operator