Functioning Hardware from Functional Specifications

Stephen A. Edwards

Columbia University

DIMACS Workshop, July 22, 2014

Where's my 10 GHz processor?

Moore's Law: Transistors Shrink and Get Cheaper



"The complexity for minimum component costs has increased at a rate of roughly a factor of two per year."

Closer to every 24 months

Gordon Moore, Cramming More Components onto Integrated Circuits, Electronics, 38(8) April 19, 1965.

Intel CPU Trends



Sutter, The Free Lunch is Over, DDJ 2005.

Data: Intel, Wikipedia, K. Olukotun

Intel CPU Trends



Sutter, The Free Lunch is Over, DDJ 2005.

Data: Intel, Wikipedia, K. Olukotun

Pollack's Rule: Diminishing Returns for Processors



Single-threaded processor performance grows with the square root of area.

It takes $4 \times$ the transistors to give $2 \times$ the performance.

Fred J. Pollack, MICRO 1999 keynote.

Graph from Borkar, DAC 2007.

Intel CPU Trends



Sutter, The Free Lunch is Over, DDJ 2005.

Data: Intel, Wikipedia, K. Olukotun

Intel Processors to Scale



What Happened in 2005?







Core 2 Duo 2006 2 cores 291 M

Xeon E5 2012 8 cores 2.3 G

Intel CPU Trends



Sutter, The Free Lunch is Over, DDJ 2005.

Data: Intel, Wikipedia, K. Olukotun



Liquid Cooled Apple Power Mac G5





2004 CMOS 1.2 kW

Where's all that power going? What can we do about it?

Dally: Calculation Cheap; Communication Costly



"Chips are power limited and most power is spent moving data

Performance = Parallelism

Efficiency = Locality

Bill Dally's 2009 DAC Keynote, The End of Denial Architecture

Parallelism for Performance; Locality for Efficiency



Dally: "Single-thread processors are in denial about these two facts"

We need different programming paradigms and different architectures on which to run them.

Massive On-Chip Parallelism: The NVIDIA GTX Titan





The NVIDIA GTX Titan/GK110 GPU

4.5	TFLOP/s
876	MHz
250	W
7	G
561	mm ²
2688	
	4.5 876 250 7 561 2688

Memory

Size	6 Gb		
Bus width	384 bits		
Clock	1.5 GHz		
Bandwidth	288 Gb/s		
Price			
\$1000			
€740			



The Future is Wires and Memory







How best to use all those transistors?

Dark Silicon



Taylor and Swanson's Conservation Cores



Custom datapaths, controllers for loop kernels; uses existing memory hierarchy

Swanson, Taylor, et al. Conservation Cores. ASPLOS 2010.

Bacon et al.'s Liquid Metal



Fig. 2. Block level diagram of DES and Lime code snippet

JITting Lime (Java-like, side-effect-free, streaming) to FPGAs Huang, Hormati, Bacon, and Rabbah, *Liquid Metal*, ECOOP 2008.

Arvind, Hoe, et al.'s Bluespec

 $\begin{array}{l} GCD \ Mod \ Rule \\ & \mathsf{Gcd}(a,b) \ if \ (a \geq b) \land (b \neq 0) \to \mathsf{Gcd}(a-b,b) \\ GCD \ Flip \ Rule \\ & \mathsf{Gcd}(a,b) \ if \ a < b \to \mathsf{Gcd}(b,a) \end{array}$



Figure 1.3 Circuit for computing Gcd(a, b) from Example 1.

Guarded commands and functions to synchronous logic Hoe and Arvind, *Term Rewriting*, VLSI 1999

Kuper et al.'s $C\lambda aSH$



Fig. 6. 4-taps FIR Filter

More operational Haskell specifications of regular structures

Baaij, Kooijman, Kuper, Boeijink, and Gerards. $C\lambda ash$, DSD 2010

What am I doing about it?















Why Functional Specifications?

Referential

transparency/side-effect freedom make formal reasoning about programs vastly easier

- Inherently concurrent and race-free (Thank Church and Rosser). If you want races and deadlocks, you need to add constructs.
- Immutable data structures makes it vastly easier to reason about memory in the presence of concurrency



Why FPGAs?

- We do not know the structure of future memory systems Homogeneous/Heterogeneous? Levels of Hierarchy? Communication Mechanisms?
- We do not know the architecture of future multi-cores Programmable in Assembly/C? Single- or multi-threaded?





Use FPGAs as a surrogate. Ultimately too flexible, but representative of the long-term solution.

A High-End FPGA: Altera's Stratix V

2500 dual-ported 2.5KB 600 MHz memory blocks; 6 Mb total 350 36-bit 500 MHz DSP blocks (MAC-oriented datapaths) 300000 6-input LUTs; 28 nm feature size



The Practical Question

How do we synthesize hardware from pure functional languages for FPGAs?

Control and datapath are easy; the memory system is interesting.

To Implement Real Algorithms, We Need

Structured, recursive data types

Recursion to handle recursive data types

Memories

Memory Hierarchy









The Type System: Algebraic Data Types

Types are primitive (Boolean, Integer, etc.) or other ADTs:

type ::= TypePrimitive type| Constr Type* | ··· | Constr Type*Tagged union

Subsume C structs, unions, and enums

Comparable power to C++ objects with virtual methods

"Algebraic" because they are sum-of-product types.

The Type System: Algebraic Data Types

Types are primitive (Boolean, Integer, etc.) or other ADTs:

type ::= TypePrimitive type| Constr Type* | ··· | Constr Type*Tagged union

Examples:

data	Intlist	=	Nil		Linked	list of	integers
			Cons Int	Intlist			

data Bintree = Leaf Int -- Binary tree of integers | Branch Bintree Bintree

data Expr = Literal Int -- Arithmetic expression | Var String | Binop Expr Op Expr

data Op = Add | Sub | Mult | Div

Example: Huffman Decoder in Haskell

```
data HTree = Branch HTree HTree
           | Leaf Char
decode :: HTree \rightarrow [Bool] \rightarrow [Char]
decode table str = bit str table
 where
   bit (False:xs) (Branch | ) = bit xs | --0: left
   bit (True:xs) (Branch r) = bit xs r --1: right
   bit x (Leaf c) = c : bit x table -- leaf
   bit []
                                              -- done
                             = []
```

Three data types:

Input bitstream Output character stream Huffman tree

```
[Bool] (list of Booleans)
[Char] (list of Characters)
HTree
```

Encoding the Types

Huffman tree nodes: (19 bits)



Boolean input stream: (14 bits)



Character output stream: (19 bits)





Optimizations



Optimizations







Hardware Synthesis: Semantics-preserving steps to a low-level dialect

Removing Recursion: The Fib Example



Transform to Continuation-Passing Style

fibk	n k	= case n (1 2 n	of → k 1 → k 1 → fibk (n−1) (λn1 →
			fibk (n–2) (λ n2 $ ightarrow$ k (n1 + n2)))
fib	n	=	fibk n ($\lambda x \rightarrow x$)

Name Lambda Expressions (Lambda Lifting)

fibk	n k = case n	of
	1 -	→ k 1
	2 -	→ k 1
	n -	ightarrow fibk (n–1) (k1 n k)
k1	n k n1 =	fibk (n–2) (k2 n1 k)
k2	n1 k n2 =	k (n1 + n2)
k0	x =	x
fib	n =	fibk n k0

Represent Continuations with a Type

```
data Cont = K0 | K1 Int Cont | K2 Int Cont
fibk n k = case (n,k) of
               (1, k) \rightarrow kk k 1
               (2, k) \rightarrow kk k 1
               (n, k) \rightarrow fibk (n-1) (K1 n k)
kk k a = case (k, a) of
      ((K1 n k), n1) \rightarrow fibk (n-2) (K2 n1 k)
      ((K2 n1 k), n2) \rightarrow kk k (n1 + n2)
      (K0, x) \rightarrow x
fib n
       =
                         fibk n K0
```

Merge Functions

```
data Cont = K0 | K1 Int Cont | K2 Int Cont
data Call = Fibk Int Cont | KK Cont Int
fibk z = case z of
    (Fibk 1 k) \rightarrow fibk (KK k 1)
    (Fibk 2 k) \rightarrow fibk (KK k 1)
    (Fibk n k) \rightarrow fibk (Fibk (n-1) (K1 n k))
    (KK (K1 n k) n1) \rightarrow fibk (Fibk (n-2) (K2 n1 k))
    (KK (K2 n1 k) n2) \rightarrow fibk (KK k (n1 + n2))
    (KK K0 x ) \rightarrow x
         =
                         fibk (Fibk n K0)
fib n
```

Add Explicit Memory Operations

```
load :: CRef \rightarrow Cont
store :: Cont \rightarrow CRef
data Cont = K0 | K1 Int CRef | K2 Int CRef
data Call = Fibk Int CRef | KK Cont Int
fibk z = case z of
    (Fibk
                 1 k) \rightarrow fibk (KK (load k) 1)
    (Fibk 2 k) \rightarrow fibk (KK (load k) 1)
    (Fibk
                 n k \rightarrow fibk (Fibk (n-1) (store (K1 n k)))
    (KK (K1 n k) n1) \rightarrow fibk (Fibk (n-2) (store (K2 n1 k)))
    (KK (K2 n1 k) n2) \rightarrow fibk (KK (load k) (n1 + n2))
    (KK K0 x) \rightarrow x
          =
                          fibk (Fibk n (store K0))
fib n
```

Syntax-Directed Translation to Hardware



Duplication Can Increase Parallelism

fib 0 = 0fib 1 = 1fib n = fib (n-1) + fib (n-2)

Duplication Can Increase Parallelism

fib 0 = 0fib 1 = 1fib n = fib (n-1) + fib (n-2) After duplicating functions:

fib 0 = 0fib 1 = 1fib n = fib' (n-1) + fib'' (n-2)fib' 0 = 0fib' 1 = 1fib' n = fib' (n-1) + fib' (n-2)fib'' 0 = 0fib'' 1 = 1fib'' n = fib'' (n-1) + fib'' (n-2)

Here, *fib'* and *fib"* may run in parallel.

Unrolling Recursive Data Structures

Original Huffman tree type:

data Htree = Branch Htree HTree | Leaf Char

Unrolled Huffman tree type:

data Htree = Branch Htree' HTree' | Leaf Char data Htree' = Branch' Htree'' HTree'' | Leaf' Char data Htree'' = Branch'' Htree HTree | Leaf'' Char

Increases locality: larger data blocks.

A type-aware cache line



 But we hit a power wall in 2005. Massive parallelism now mandatory

Communication is the culprit



 Dark Silicon is the future: faster transistors; most must remain off

 Custom accelerators are the future; many approaches

 My project: A Pure Functional Language to FPGAs



Algebraic Data Types in Hardware

Optimizations

Removing recursion

Encoding the Types

Huffman tree nodes: (19 bits)



Boolean input stream: (14 bits)



Character output stream: (19 bits)



Syntax-Directed Translation to Hardware

