



Building Zynq Accelerators with Vivado High Level Synthesis

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Schedule

- Motivation for Zynq and HLS (5 min)
- Zynq Overview (45 min)
- HLS training (the condensed version) (1.5 hours)
- Zynq Systems with HLS (45 min)

Motivation

► ASICs *are* being displaced by programmable platforms

- Packaging, verification costs dominate
- FPGA/ASSP process advantage over commodity ASIC process
- Full-/semi-custom design vs. standard cell ASIC

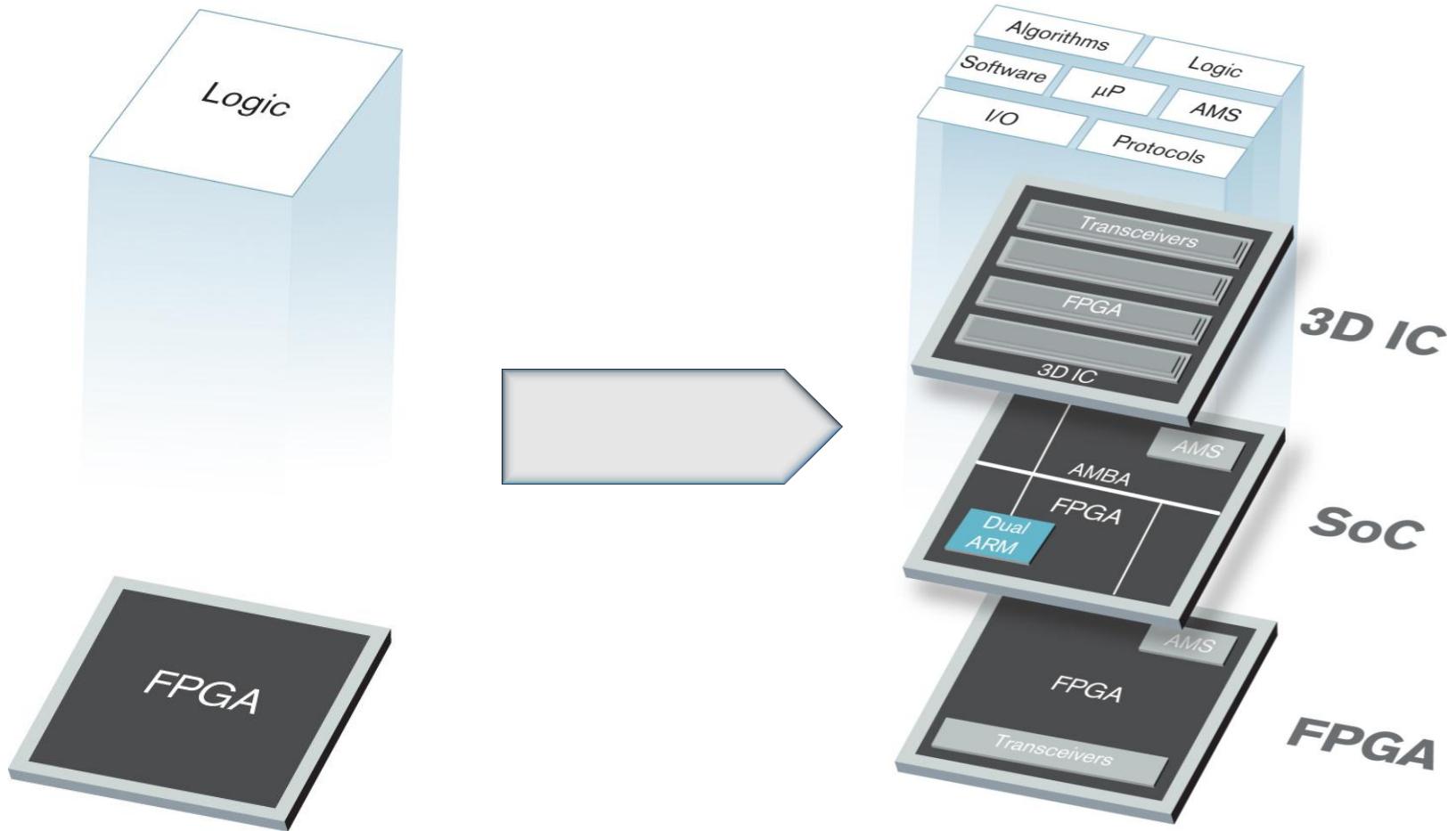
► Lots of competing programmable platforms

- CPU+GPGPU
- CPU+DSP+hard accelerators (e.g. OMAP, Davinci, etc.)
- Multicore
- FPGAs

► From FPGAs to “All Programmable Devices”

- ‘Small’ devices are very capable with increasing integration
- ‘Big’ devices are getting REALLY big.

Xilinx Technology Evolution



➤ **Programmable Logic Devices**
Enables Programmable “Logic”

➤ **All Programmable Devices**
Enables Programmable “Systems Integration”

Zynq-7000 Family Highlights

► Complete ARM®-based Processing System

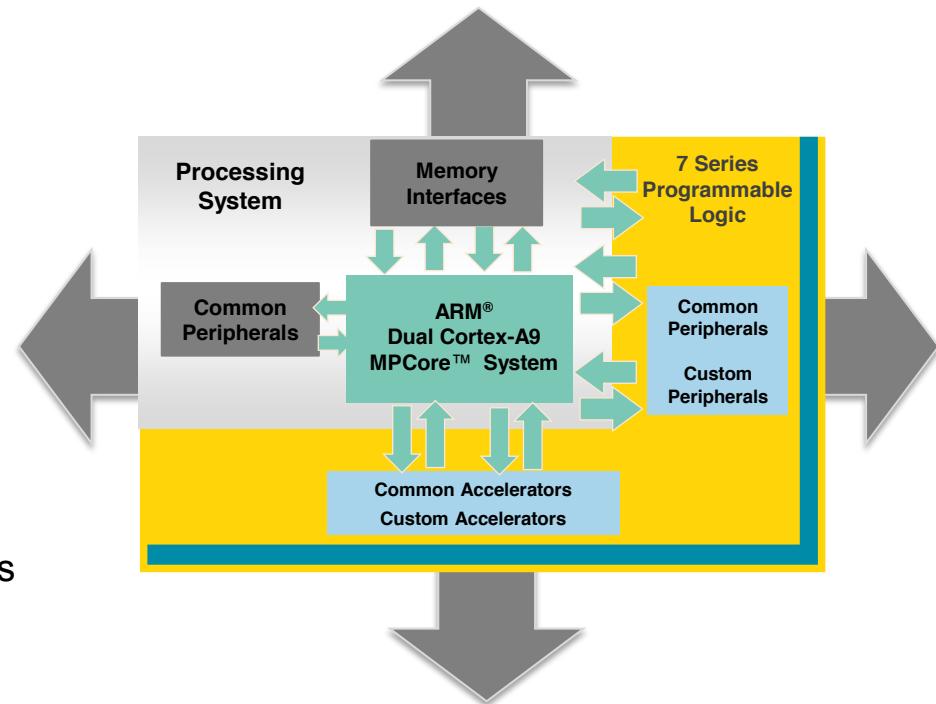
- Dual ARM Cortex™-A9 MPCore™, processor centric
- Integrated memory controllers & peripherals
- Fully autonomous to the Programmable Logic

► Tightly Integrated Programmable Logic

- Used to extend Processing System
- High performance ARM AXI interfaces
- Scalable density and performance

► Flexible Array of I/O

- Wide range of external multi-standard I/O
- High performance integrated serial transceivers
- Analog-to-Digital Converter inputs



Zynq-7000 AP SoC Applications Mapping

CLUSTER	MARKET	KEY APPLICATIONS	SAME PROCESSING SYSTEM Different Programmable Logic Densities			
			<ul style="list-style-type: none"> ● Optimal for application 			
			Z-7010	Z-7020	Z-7030	Z-7045
Intelligent Video	Auto	Driver Assistance, Driver Info, Infotainment	●	●		
	Consumer	Business-class Multi-function Printers	●	●	●	
	ISM	IP & Smart Cameras	●	●	●	
		Medical Diagnostics, Monitoring and Therapy	●	●		
		Medical Imaging	●		●	●
	Broadcast	Prosumer / Studio Cameras, Transcoders		●	●	●
Comms	A&D	Video / Night Vision Equipment	●	●		
	A&D	Milcomms, Cockpit & Instrumentation		●	●	●
	Wireless	LTE Radio, Baseband, Enterprise Femto		●	●	●
Control	Wired	Routers, Switches, Multiplexers, Edge Cards			●	●
	ISM	Motor Control and Programmable Logic Controller (PLC)	●	●	●	
	A&D	Missiles, Smart Munitions			●	●
Bridging	Broadcast	EdgeQAMs, Routers, Switchers, Encoders / Decoders			●	●
	ISM	Industrial Networking	●	●	●	

← 10X →

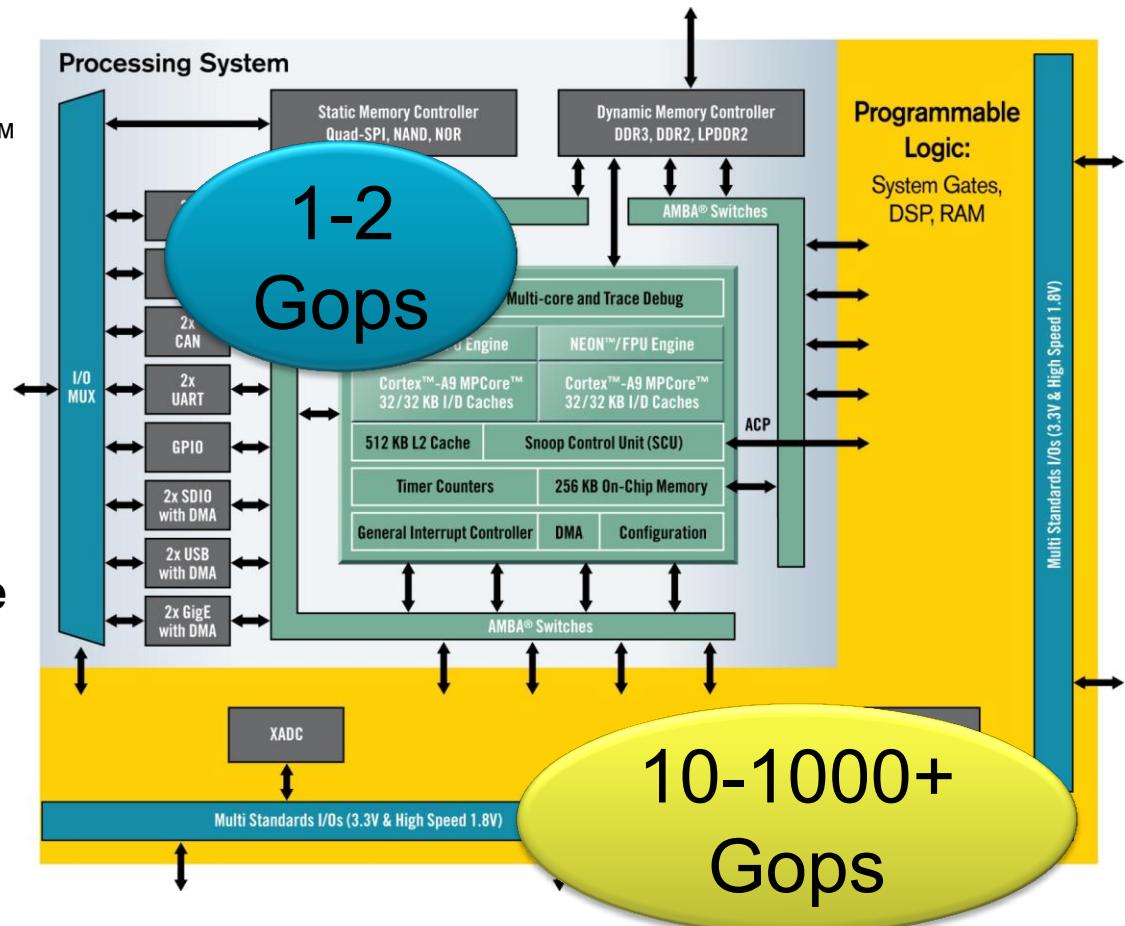
Zynq-7000 Embedded Processing Platform

Processor core complex

- Two ARM® Cortex™-A9 with NEON™ extensions
- Floating Point support
- Up to 1 GHz operation
- L2 Cache – 512KB Unified
- On-Chip Memory of 256KB
- Integrated Memory Controllers
- Run full Linux

State-of-the-art programmable logic

- 28K-235K logic cells
- High bandwidth AMBA interconnect
- ACP port - cache coherency for additional soft processors



How to Leverage the Compute Power of the Fabric?

Systems in FPGA: 3 independent pieces

► Interface IP blocks (HDMI, Memory Controller)

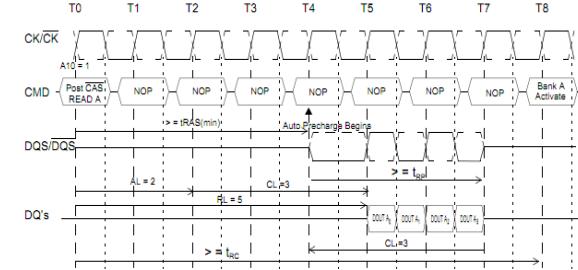
- Everything within 1 or 2 cycles of IO, "glue logic"
- Timing accurate
- Structural RTL + constraints, Spice, IBIS models

► Core IP (microblaze, NOC)

- Cycle accurate
- Structural or synthesizable RTL

► Application-specific IP

- Differentiation/added value
- High level throughput/latency constraints
- Synthesizable RTL or Algorithmic spec



	cycle 1	cycle 2	cycle 3	cycle 4	cycle 5	cycle 6	cycle 7	cycle 8	cycle 9
instruction 1	IF	OF	EX	MEM	WB				
instruction 2		IF	OF	EX	MEM	MEM	MEM	WB	
instruction 3			IF	OF	EX	Stall	Stall	MEM	WB

▫ Summary of overall latency (clock cycles)

- Best-case latency: 1
- ◆ Average-case latency: 1202026
- Worst-case latency: 4501354

▫ Summary of loop latency (clock cycles)

▫ Loop 1

- # Trip count: 0 ~ 2047
- ⌚ Latency: 0 ~ 4501353

High Level Synthesis

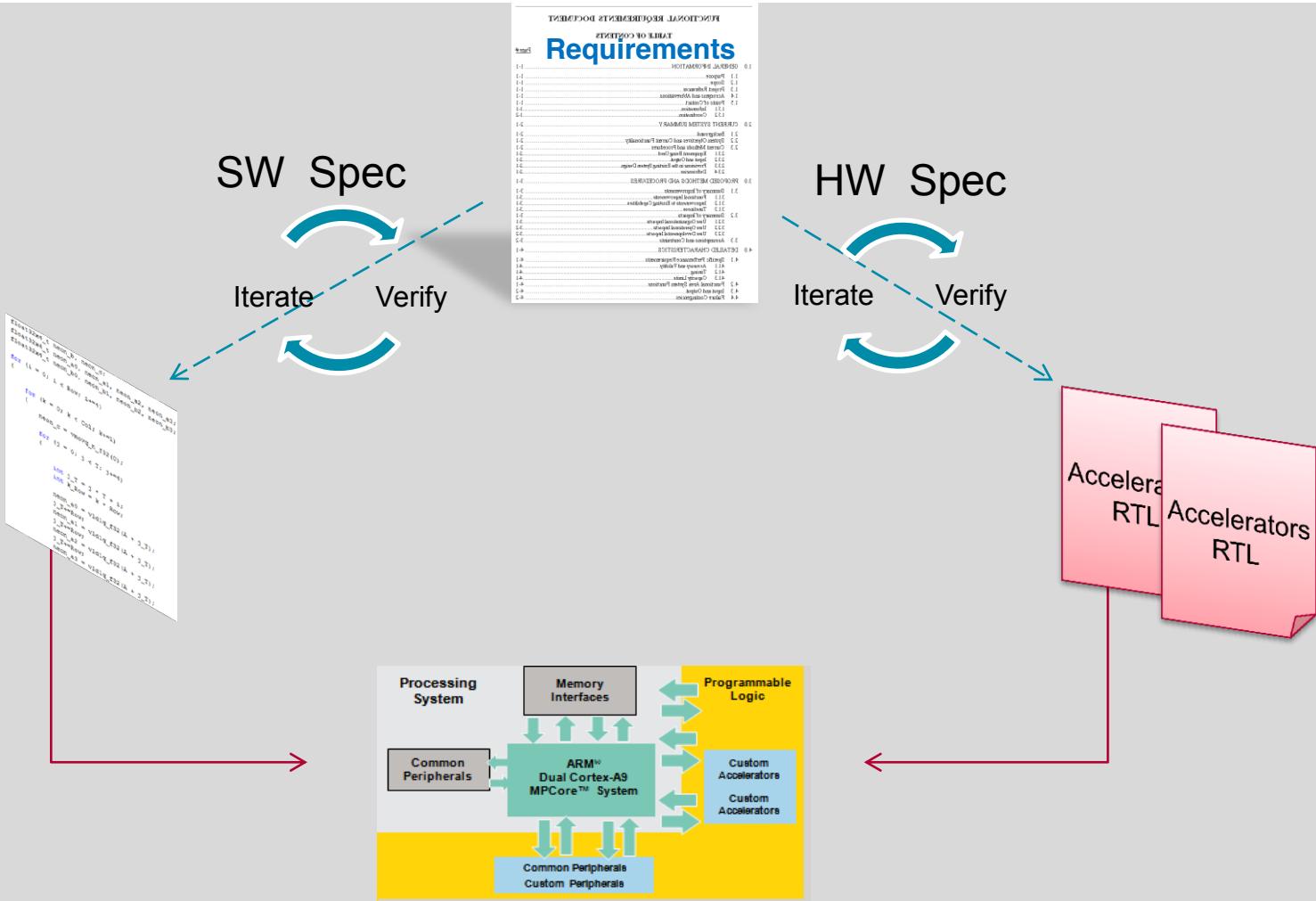
► Generating Application-Specific IP from Algorithmic C specification

- Focus on Macro-architecture exploration... leave microarchitecture to tool

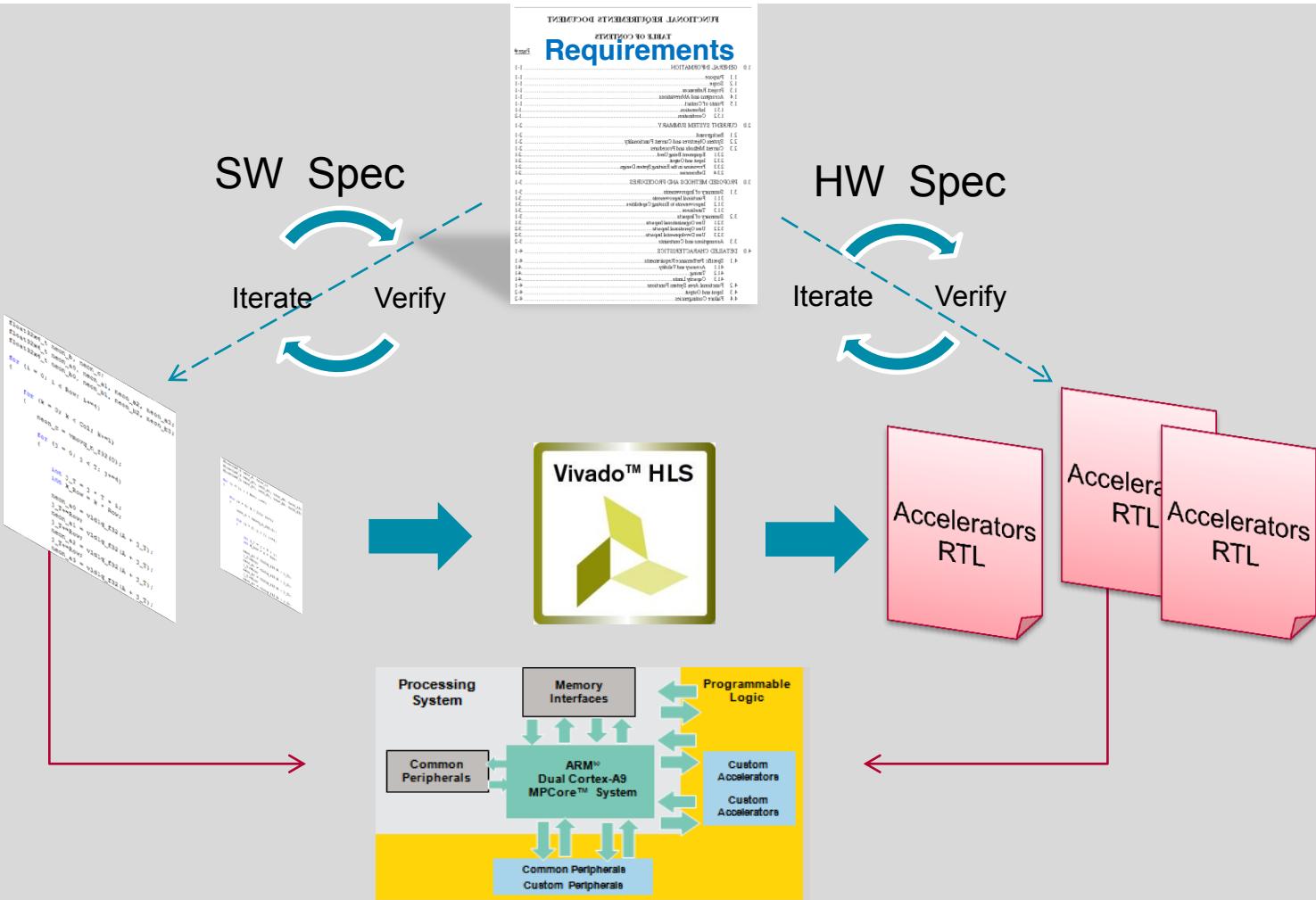
► A few key problems

- Extracting lots of parallelism
 - Statically scheduled Instruction-level parallelism (in loops)
 - Dynamically controlled task-level parallelism (between loops)
- Analyzing pointer aliases
 - Most arrays map into BRAM, rather than global address space
- Understanding performance
 - Good timing models for FPGA synthesis
 - Interval/Latency analysis

All Programmable SOC Approach



Vivado High-Level Synthesis



Accelerates Algorithmic C to Co-Processing Accelerator Integration

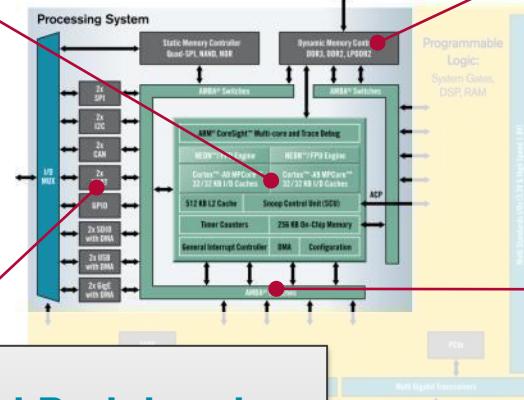


Zynq Overview

Complete ARM-based Processing System

Processor Core Complex

- Dual ARM Cortex-A9 MPCore with NEON™ extensions
- Single / Double Precision Floating Point support
- Up to 1 GHz operation



Integrated Memory Mapped Peripherals

- 2x USB 2.0 (OTG) w/DMA
- 2x Tri-mode Gigabit Ethernet w/DMA
- 2x SD/SDIO w/DMA
- 2x UART, 2x CAN 2.0B, 2x I2C, 2x SPI, 32b GPIO

High BW Memory

- Internal
 - L1 Cache – 32KB/32KB (per Core)
 - L2 Cache – 512KB Unified
- On-Chip Memory of 256KB
- Integrated Memory Controllers (DDR3, DDR2, LPDDR2, 2xQSPI, NOR, NAND Flash)

AMBA Open Standard Interconnect

- High bandwidth interconnect between Processing System and Programmable Logic
- ACP port for enhanced hardware acceleration and cache coherency for additional soft processors

➤ Processing System Ready to Program

Powerful Application Processor at Heart

The Application Processor Unit (APU)

► Dual ARM Cortex-A9 MPCore with NEON extensions

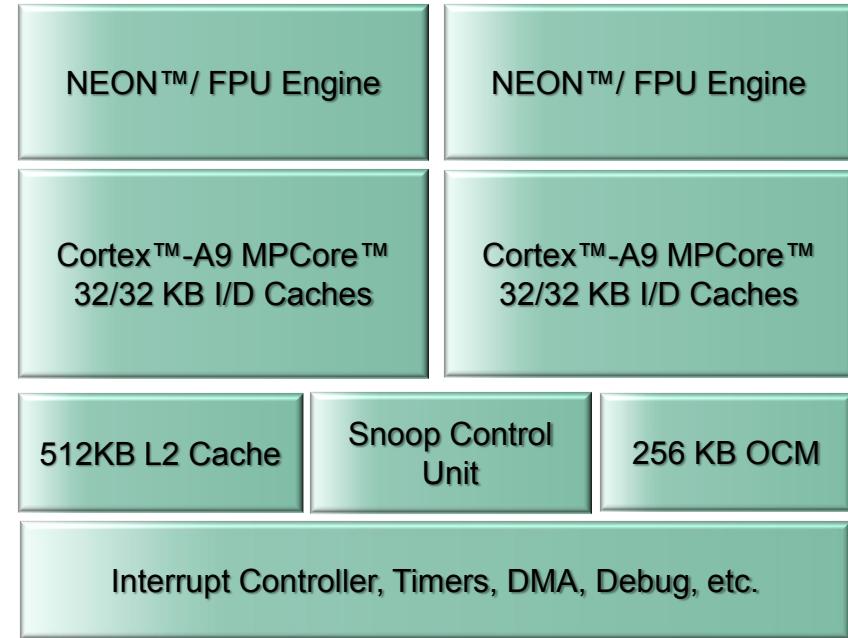
- Up to 1 GHz operation (7030 & 7045)
- Multi-issue (up to 4), Out-of-order, Speculative
- Separate 32KB Instruction and Data Caches with Parity

► Snoop Control Unit

- L1 Cache Snoop Control
 - Snoop filtering monitors cache traffic
 - Accelerator Coherency Port

► Level 2 Cache and Controller

- Shared 512 KB Cache with parity
- Lockable



► On-Chip Memory (OCM)

- Dual-ported 256KB
- Low-latency CPU access
- Accessible by DMAs, Programmable Logic, etc.

Processing System External Memories

Built-in Controllers and dedicated DDR Pins

➤ DDR controller

- DDR3 @ up to DDR1333
- DDR2 @ up to DDR800
- LPDDR2 @ up to DDR800
- 16 bit or 32 bit wide; ECC on 16 bit
- 73 dedicated DDR pins

➤ Non-volatile memory (processor boot and FPGA configuration)

- NAND flash Controller (8 or 16 bit w/ ECC)
- NOR flash/SRAM Controller (8 bit)
- Quad SPI (QSPI) Controller

Zynq OS Boot process

► Multi-stage boot process

- Stage 0: Runs from ROM
 - loads FSBL from boot device to OCM
- Stage 1 (FSBL): Runs from OCM
 - loads Uboot from boot device to DDRx memory
 - Initiates PS boot and PL configuration
- Stage 2 (e.g. Uboot): runs from DDR
 - loads Linux kernel, initial ramdisk, and device tree from any location
 - May access FPGA
- OS boot (e.g. Linux): runs from DDR

► Supports ‘secure boot’ chain of trust

Typical Linux Boot from SD card

► Typical boot image (contents of BOOT.BIN)

```
the_ROM_image:  
{  
    [bootloader] zynq_fsbl.elf  
    system.bit  
    u-boot.elf  
}
```

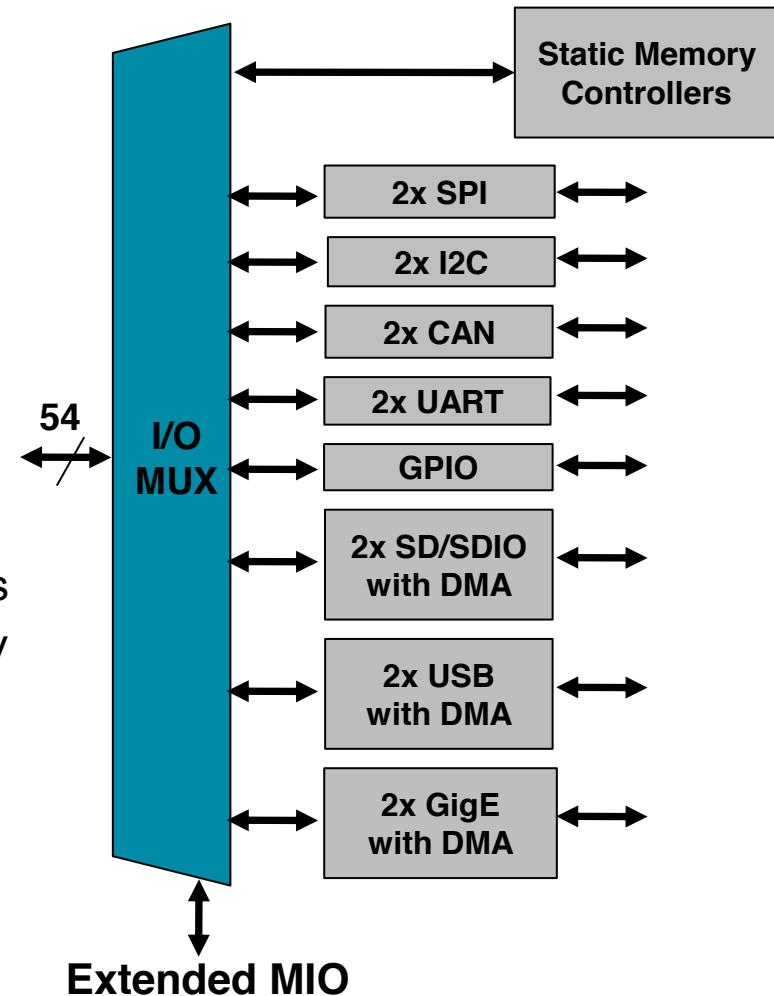
► Typical SD card contents

```
zynq> ls  
devicetree.dtb  
BOOT.bin  
ramdisk8M.image.gz  
uImage
```

Comprehensive set of Built-in Peripherals

Enabling a wide set of IO functions

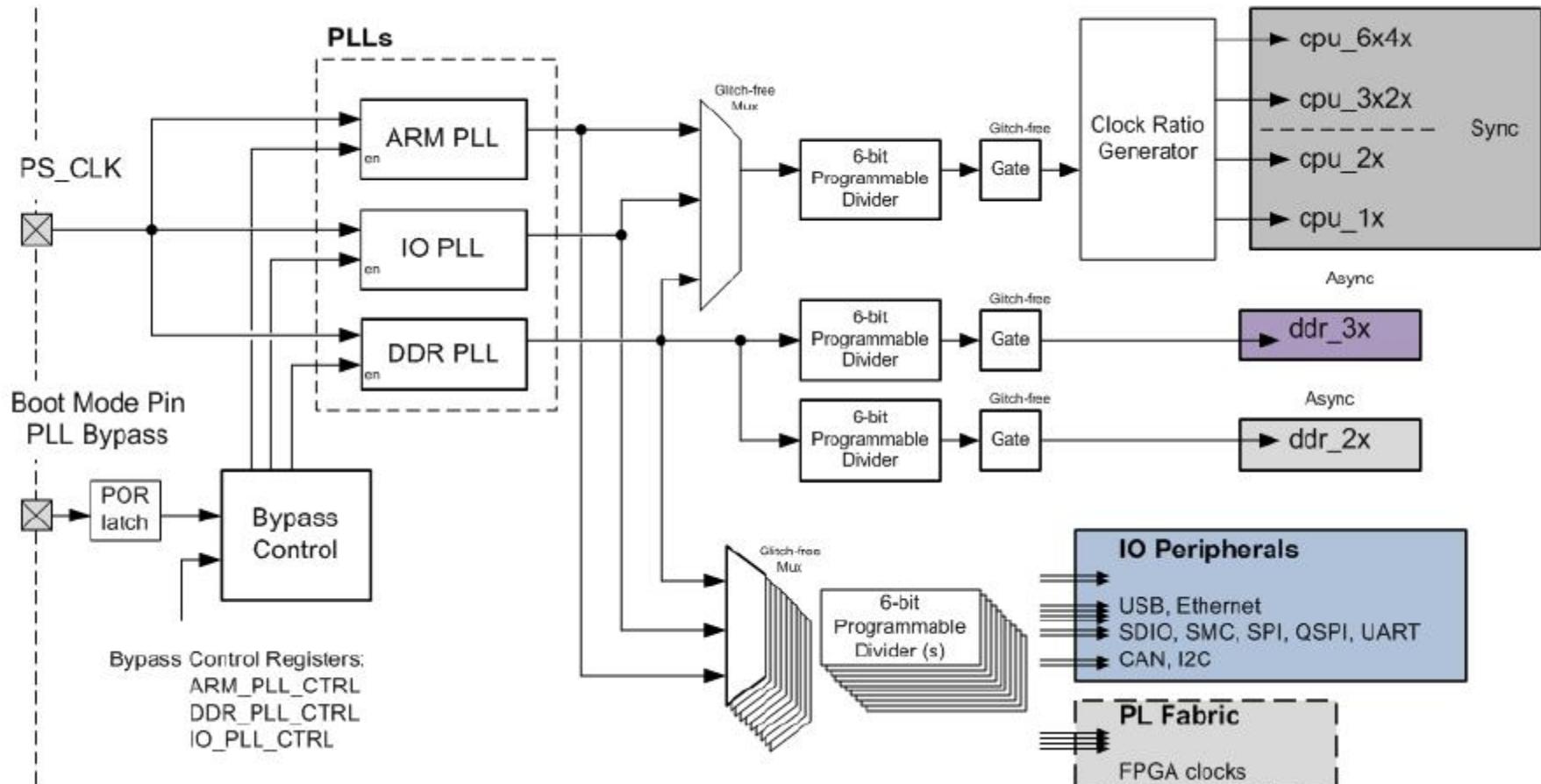
- Two USB 2.0 OTG/Device/Host
- Two Tri- Mode GigE (10/100/1000)
- Two SD/SDIO interfaces
- Two CAN 2.0B, SPI , I2C , UART
- Four GPIO 32bit Blocks
- Multiplexed Input/Output (MIO)
 - Multiplexed output of peripheral and static memories
 - Two I/O Banks: each selectable - 1.8V, 2.5V or 3.3V
- Extended MIO
 - Enables use of Select IO with PS peripherals
 - FPGA must be configured before using EMIO connections
 - EMIO connections use FPGA routing



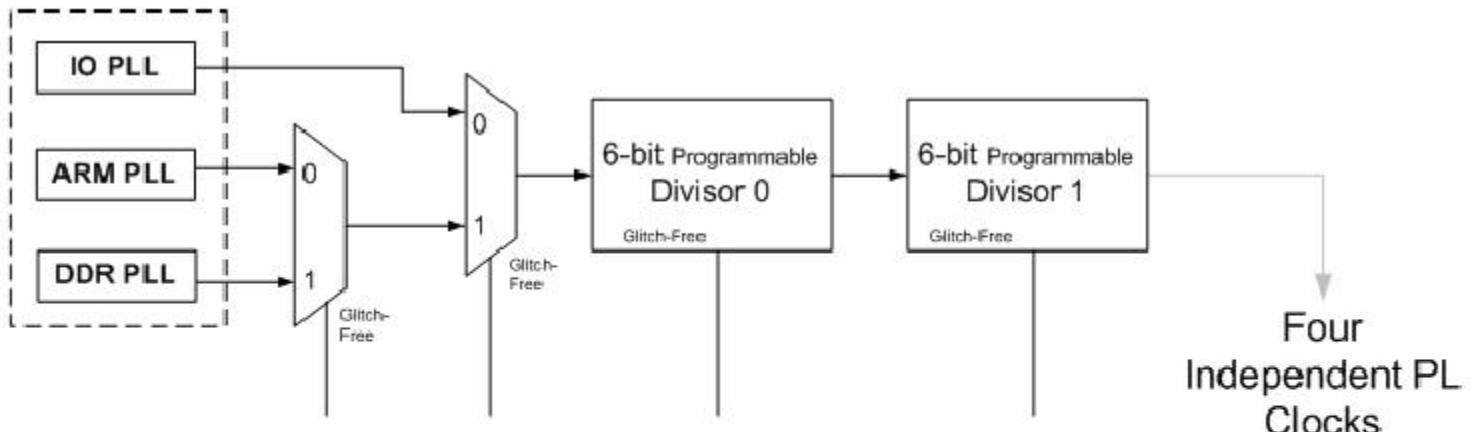
Multiplexed I/O (MIO) Pinout

IP	MIO	Extendable MIO in Programmable Logic
QSPI NOR/SRAM NAND	Yes	No
USB:0,1	Yes, Phy off chip	No
SDIO:0,1	Yes – 50MHz	Yes – 25MHz
SPI:0,1 I2C:0,1 CAN:0,1 GPIO	Yes	Yes
GigE:0,1	RGMII v2.0 (HSTL) Phy off chip	Supports GMII, RGMII v2.0 (HSTL), RGMII v1.3 (LVCMOS), RMII, MII, SGMII with wrapper in Programmable Logic
UART:0,1	Simple UART: Only 2 pins (Tx & Rx)	Full UART (Tx, Rx, DTR, DCD, DSR, RI, RTS & CTS) either require: <ul style="list-style-type: none">• 2 Processing System pins (Rx & Tx) through MIO + 6 additional Programmable Logic pins• 8 Programmable Logic pins

Clock Generator Block Diagram



Clocking the PL



PL Fabric Clock	Control Register	Mux Ctrl Field	Mux Ctrl Field	Divider 0 Ctrl Field	Divider 1 Ctrl Field
PL Fabric 0	FPGA0_CLK_CTRL	SRCSEL_4	SRCSEL_5	DIVISOR 0, 13:8	DIVISOR 1, 25:20
PL Fabric 1	FPGA1_CLK_CTRL	SRCSEL_4	SRCSEL_5	DIVISOR 0, 13:8	DIVISOR 1, 25:20
PL Fabric 2	FPGA2_CLK_CTRL	SRCSEL_4	SRCSEL_5	DIVISOR 0, 13:8	DIVISOR 1, 25:20
PL Fabric 3	FPGA3_CLK_CTRL	SRCSEL_4	SRCSEL_5	DIVISOR 0, 13:8	DIVISOR 1, 25:20

→ FCLKCLK0
→ FCLKCLK1
→ FCLKCLK2
→ FCLKCLK3

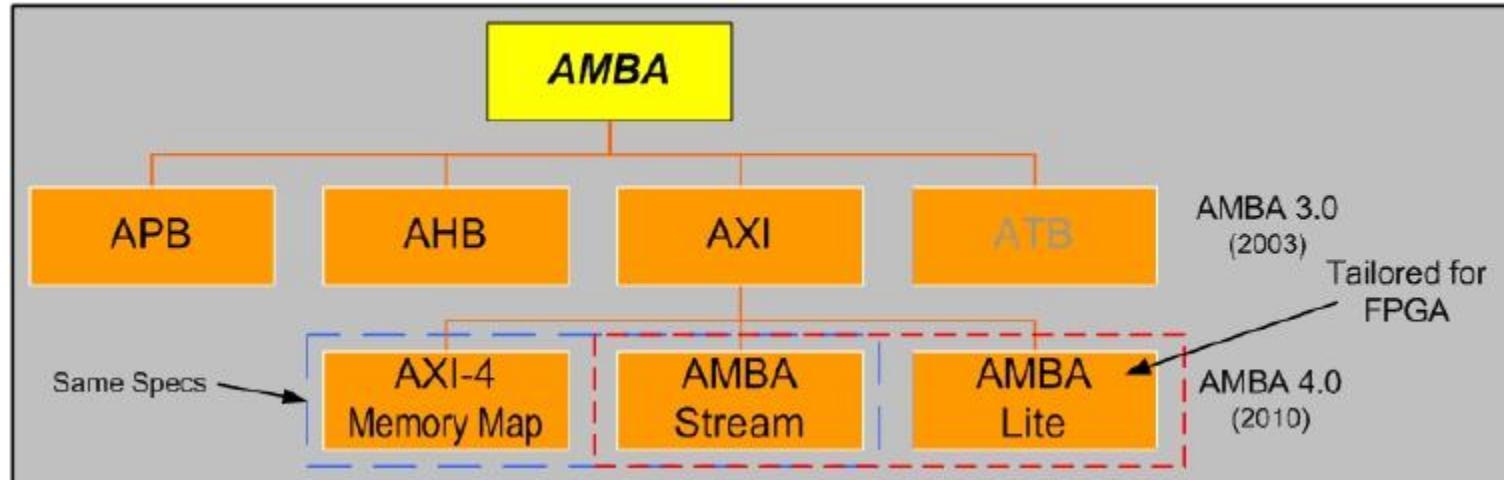
► Other features:

- associated reset (FCLKRSTn)
- software clock counter
- clock pause trigger (FCLKCLKTRIGxN)

Interrupts

- **16 peripheral interrupts from PL to PS**
 - Used for accelerators and peripherals in PL
- **4 processor-specific interrupts from PL to PS**
- **28 interrupts from PS peripherals to PL**
 - PS peripherals can be serviced from Microblaze in fabric

AXI is Part of AMBA: Advanced Microcontroller Bus Architecture



Interface	Features	SimilarTo
MemoryMap/Full	Traditional address/data burst (single address,multiple data)	PLBv46,PCI
Streaming	Dataonly,burst	LocalLink/DSP interfaces/FIFO/FSL
Lite	Traditional address/data—no burst (single address,multiple data)	PLBv46single OPB

AXI Interface: Streaming

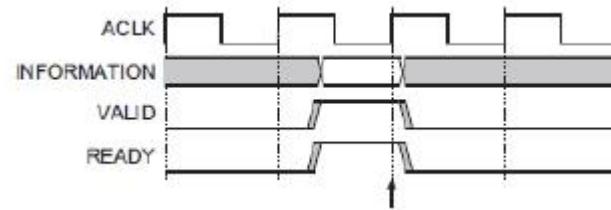
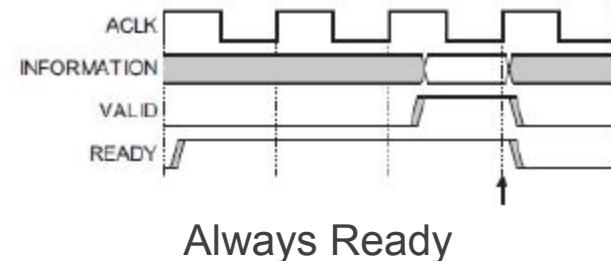
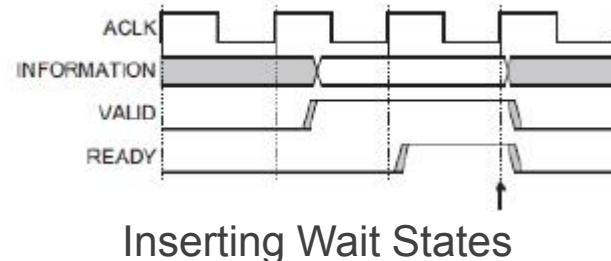
► AXI Streams are fully handshaked

- Data is transferred when source asserts VALID and destination asserts READY

► ‘Information’ includes DATA and other side channel signals

- STRB
- KEEP
- LAST
- ID
- DEST
- USER

► Most of these are optional



AXI Interface: AXI4

➤ Memory mapped interfaces consist of 5 streams

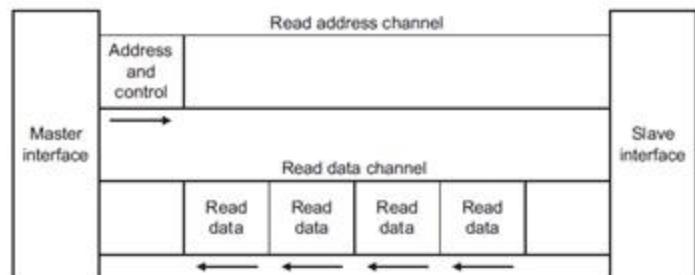
- Read Address
- Read Data
- Write Address
- Write Data
- Write Acknowledge

➤ Burst length limited to 256

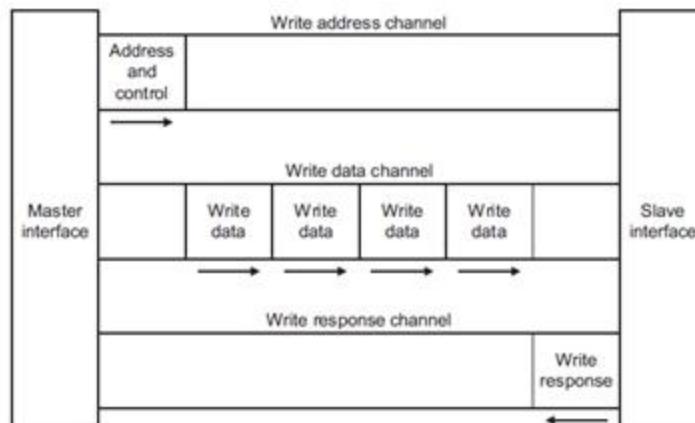
➤ Data width limited to 256 bits for Xilinx IP

➤ AXI Lite is a subset

- no bursts
- 32 bit data width only



AXI4 READ



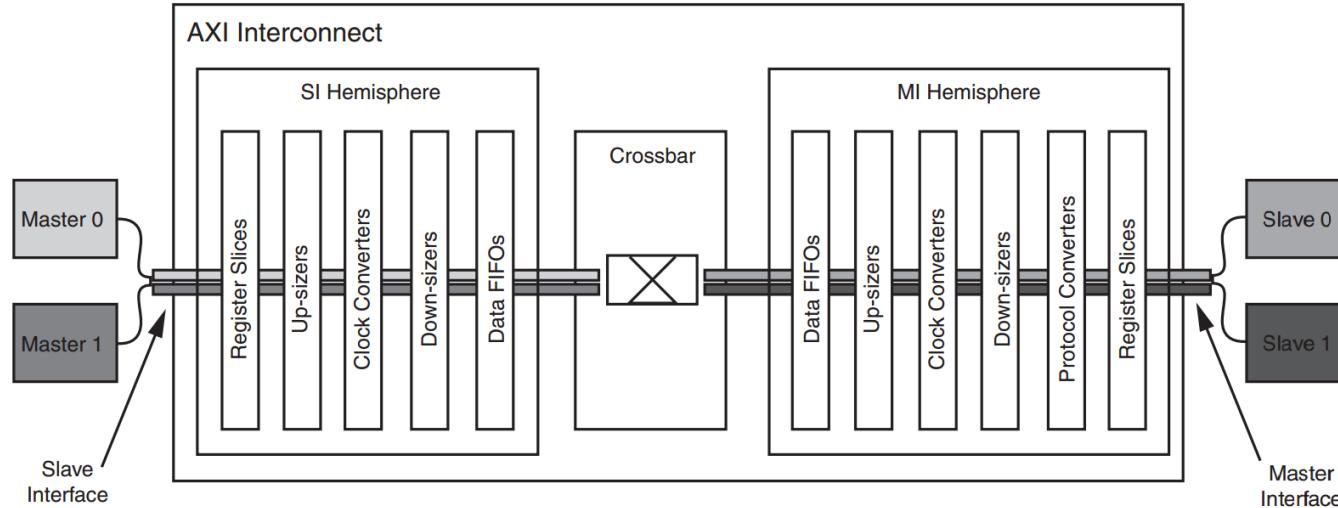
AXI4 Write

AXI Interconnect IP in PS

➤ Uses AXI4 Memory Mapped Interfaces

- Automatic width conversion
- Automatic AXI3/AXI4 Lite protocol conversion
- Automatic clock-domain crossing

- Configurable sparse crossbar or shared bus
- Optional buffering fifos
- Optional timing isolation registers

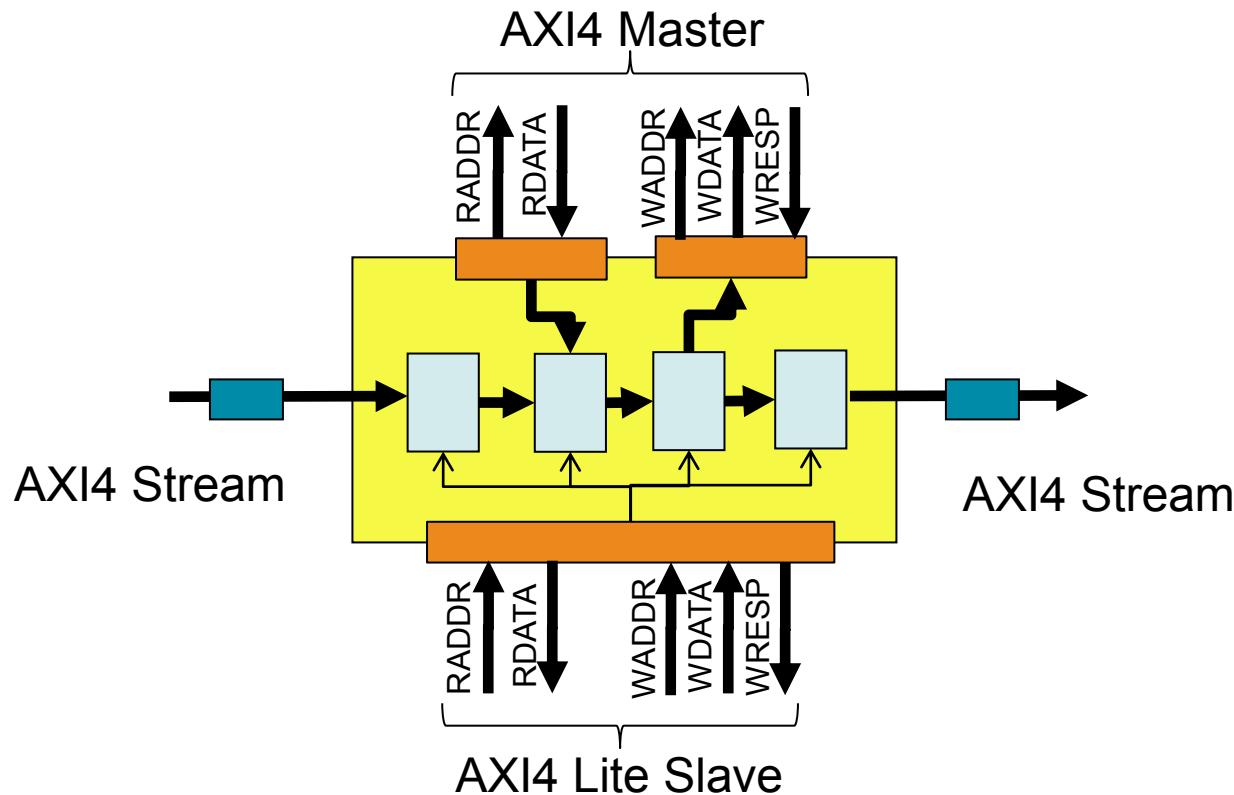


AXI Interconnect IP Details

- **Centralized arbitration with parallel data**
 - Arbitration optimized for 3+ data beats per burst
- **Buffering allows address pipelining**
 - However, Masters and Slaves have practical limits on pipelining
 - Described using Master ISSUING and Slave ACCEPTANCE parameters
 - Arbitration uses these parameters to limit head-of-line blocking

AXI based accelerators

► HLS accelerators will combine lots of AXI interfaces



Zynq AXI Interfaces

The HP logo consists of a red right-pointing arrow followed by the letters "HP" in a black sans-serif font.

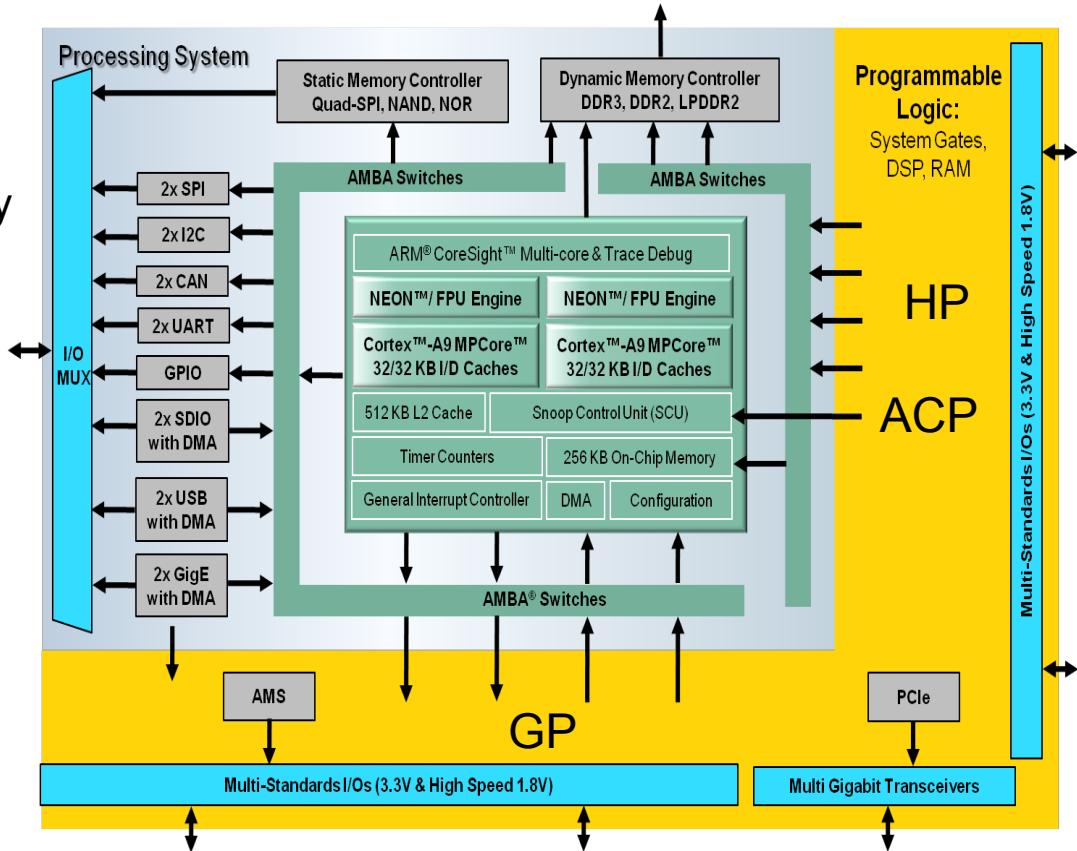
- 4 x 64 bit Slave interfaces
 - Optimized for high bandwidth access from PL to external memory

 GP

- 2 x 32 bit Slave interfaces
 - Optimized for access from PL to PS peripherals
 - 2 x 32 bit Master interfaces
 - Optimized for access from processors to PL registers

ACP

- 1 x 64 bit Slave interface
 - Optimized for access from PL to processor caches



GP Port Summary

- GP ports are designed for maximum flexibility
- Allow register access from PS to PL or PL to PS

- Good for Synchronization
- Prefer ACP or HP port for data transport

HP Port Summary

- HP ports are designed for maximum bandwidth access to external memory and OCM
- When combined can saturate external memory and OCM bandwidth
 - HP Ports : $4 * 64 \text{ bits} * 150 \text{ MHz} * 2 = 9.6 \text{ GByte/sec}$
 - external DDR: $1 * 32 \text{ bits} * 1066 \text{ MHz} * 2 = 4.3 \text{ GByte/sec}$
 - OCM : $64 \text{ bits} * 222 \text{ MHz} * 2 = 3.5 \text{ GByte/sec}$
- Optimized for large burst lengths and many outstanding transactions
- Large data buffers to amortize access latency
- Efficient upsizing/downsizing for 32 bit accesses

ACP Port Summary

► ACP allows limited support for Hardware Coherency

- Allows a PL accelerator to access cache of the Cortex-A9 processors
- PL has access to through the same path as CPUs
 - including caches, OCM, DDR, and peripherals
- Access is low latency (assuming data is in processor cache)
 - no switches in path

► ACP does not allow full coherency

- PL is not notified of changes in processor caches (different from ACE)
- Use “event bus” or register write of PL register for synchronization

► ACP is compromise between bandwidth and latency

- Optimized for cache line length transfers
- Low latency for L1/L2 hits
- Minimal buffering to hide external memory latency
- One shared 64 bit interface, limit of 8 masters

ACP Details

► Must access complete cache lines (32 bytes)

- LENGTH = 3 (i.e. 4 data beats)
- SIZE = 3 (i.e. transfer 8 bytes per data beat)
- STRB = 0xFF (i.e. all data will be read/written)
- Proper address alignment
 - Incremental burst with 32 byte alignment
 - Wrapped burst with 8 byte alignment

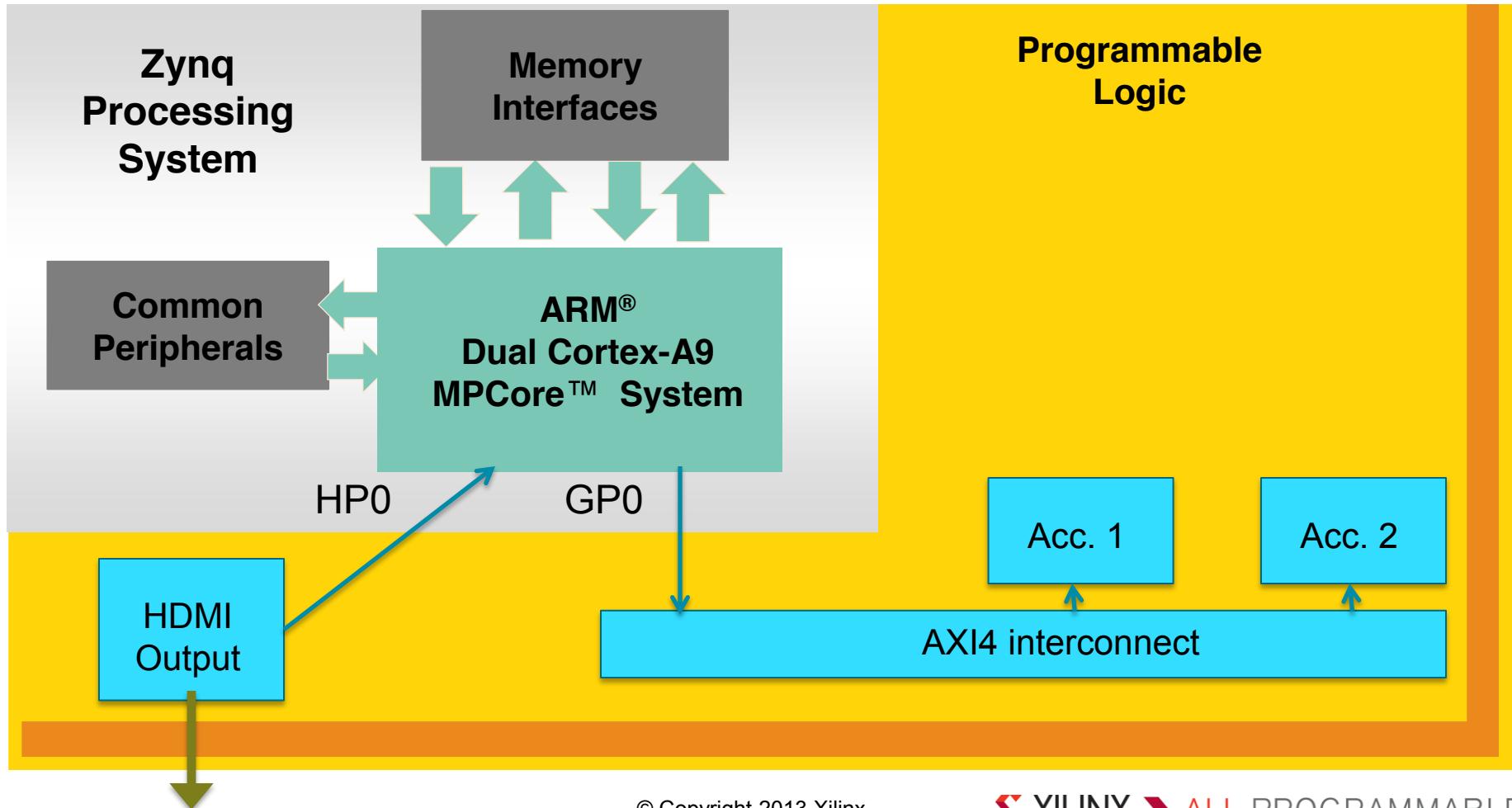
► USER[0] = 1 and CACHE[0] = 1 to hit in cache

► <http://infocenter.arm.com/help/index.jsp?topic=/com.arm.doc.ddi0407f/BABCDDIA.html>

Accelerator Architecture (With Bus Slave)

Pro: Simple System Architecture

Con: Limited communication bandwidth



Bus Slave Accelerator Communication

➤ Write to Accelerator

- processor writes to uncached memory location

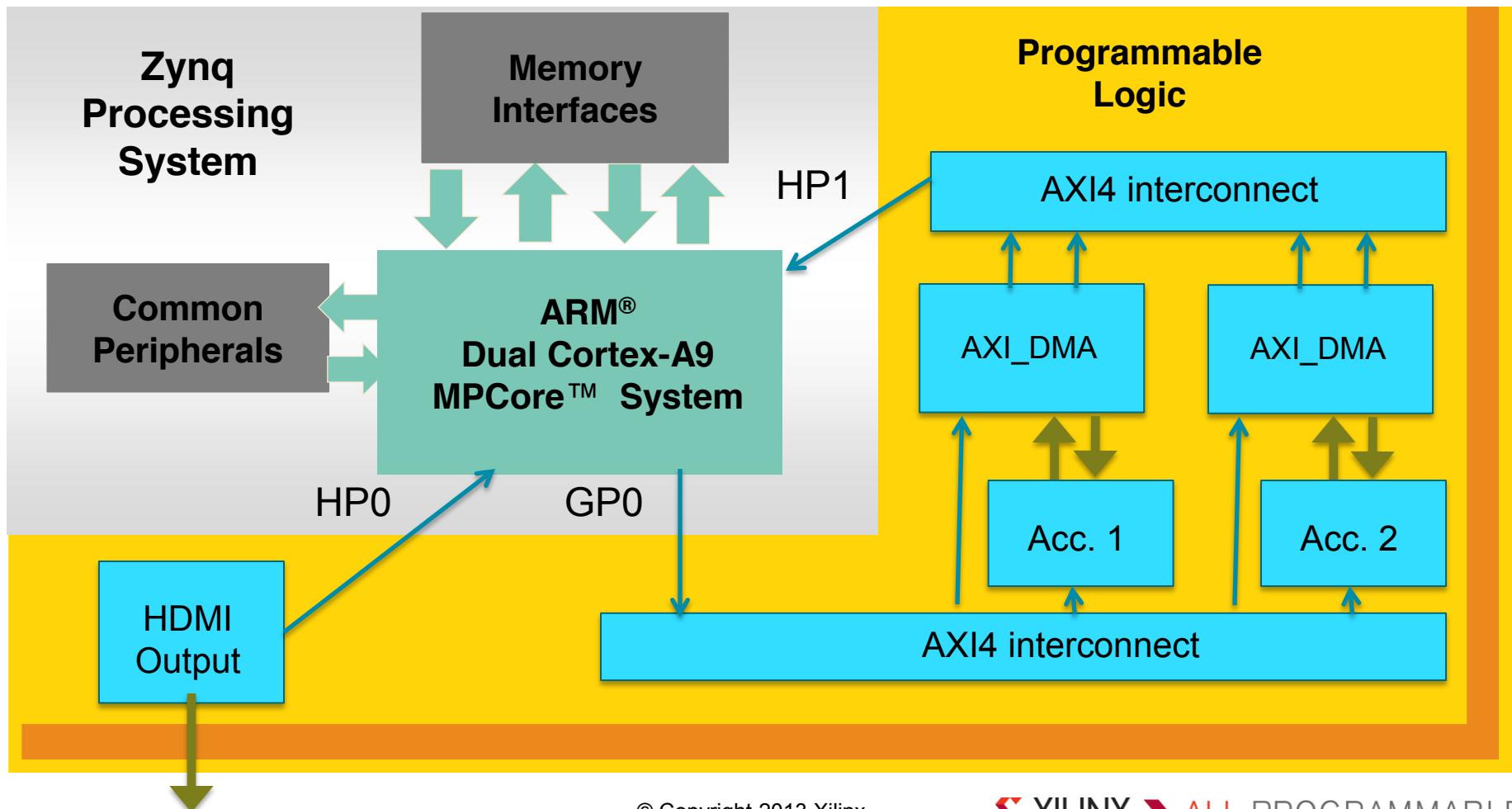
➤ Read from Accelerator

- processor reads from uncached memory location

Architecture (With DMA)

Pro: High Bandwidth Communication

Con: Complicated System Architecture, High Latency



AXI DMA-based Accelerator Communication

► Write to Accelerator

- processor allocates buffer
- processor allocates scatter-gather list
- processor initializes scatter-gather list with physically continuous segments
- processor writes data into buffer
- processor flushes cache for buffer
- processor pushes scatter-gather list to DMA register

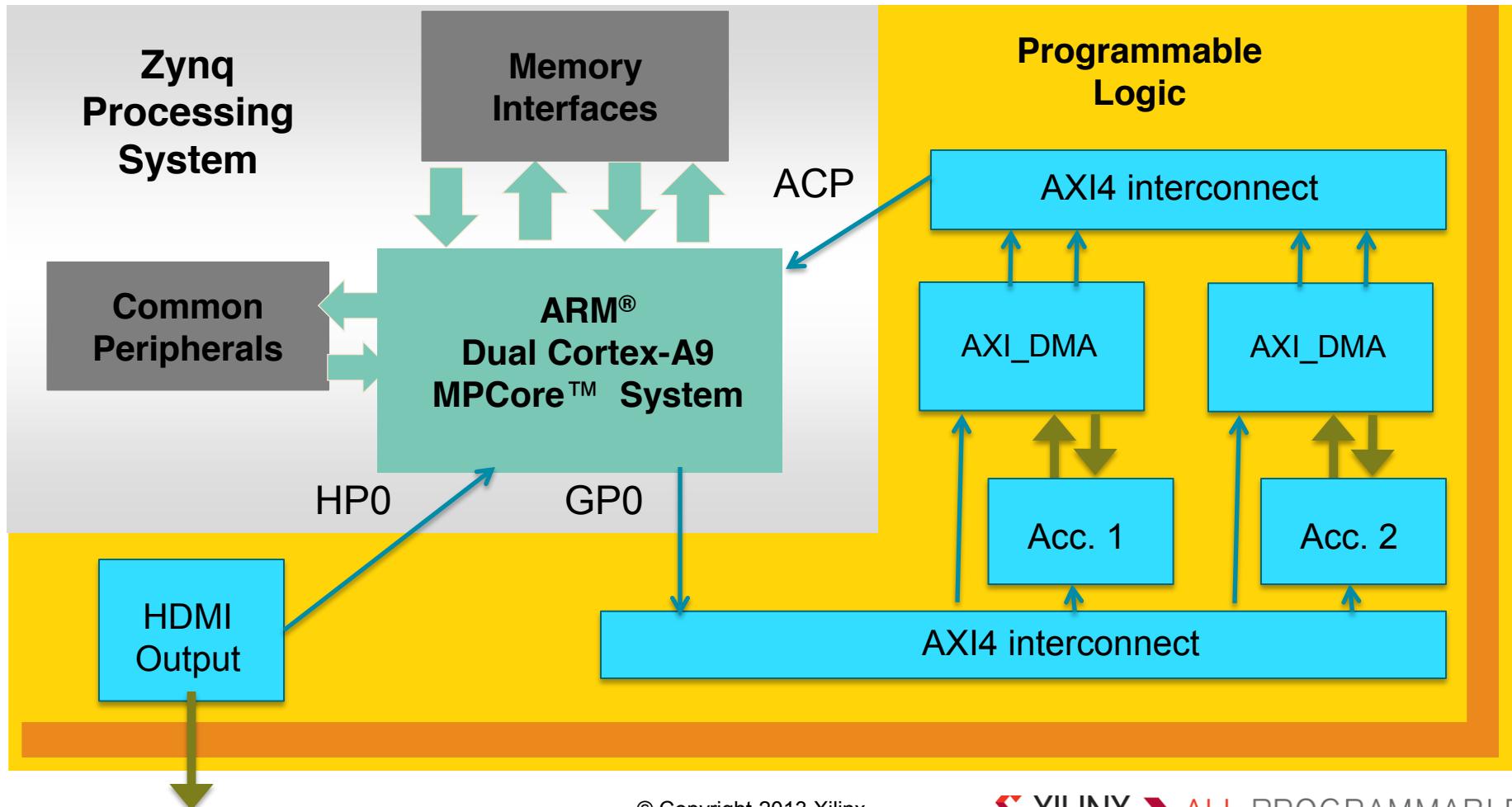
► Read from Accelerator

- processor allocates buffer
- processor allocates scatter-gather list
- processor initializes scatter-gather list with physically continuous segments
- processor pushes scatter-gather list to DMA register
- processor waits for DMA complete
- processor invalidates cache for buffer
- processor reads data from buffer

Architecture (With Coherent DMA)

Pro: Low latency, high-bandwidth communication

Con: Complicated system architecture, Limited to data that fits in caches



Coherent AXI DMA-based Accelerator Communication

► Write to Accelerator

- processor allocates buffer
- processor allocates scatter-gather list
- processor initializes scatter-gather list with physically continuous segments
- processor writes data into buffer
- ~~processor flushes cache for buffer~~
- processor pushes scatter-gather list to DMA register

► Read from Accelerator

- processor allocates buffer
- processor allocates scatter-gather list
- processor initializes scatter-gather list with physically continuous segments
- processor pushes scatter-gather list to DMA register
- processor waits for DMA complete
- ~~processor invalidates cache for buffer~~
- processor reads data from buffer



How Does HLS Work?

How Does HLS Work?

- Overview of HLS
- HLS Coding and Design Capture
- Default Behaviors
- Performance Optimization
- Area Optimization
- Interface Definition



Overview of HLS

HLS Premise: 1 C Code – Multiple HW Implementations

One body of code:
Many hardware outcomes

```
...  
loop: for (i=3;i>=0;i--) {  
    if (i==0) {  
        acc+=x*c[0];  
        shift_reg[0]=x;  
    } else {  
        shift_reg[i]=shift_reg[i-1];  
        acc+=shift_reg[i]*c[i];  
    }  
}....
```

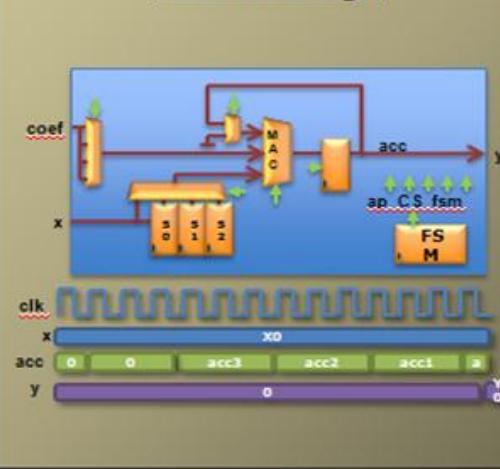
Before going into details,
let's look under the hood
....

The same hardware is used for
each iteration of the loop:
• Small area
• Long latency
• Low throughput

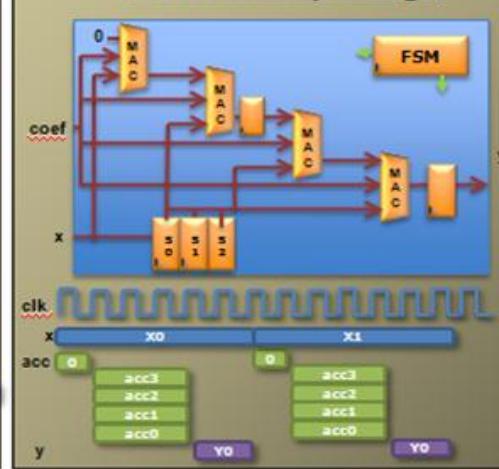
Different hardware is used for each
iteration of the loop:
• Higher area
• Short latency
• Better throughput

Different iterations are executed
concurrently:
• Higher area
• Short latency
• Best throughput

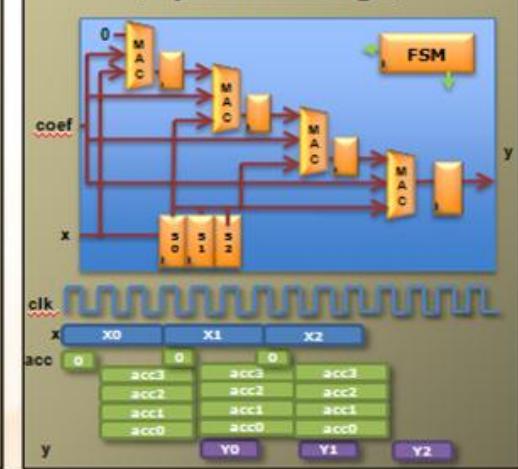
Default Design



Unrolled Loop Design



Pipelined Design



Attributes of a Program for Synthesis

► Functions

- Functions define hierarchy and control regions

► Function Parameters:

- Define the RTL I/O Ports

► Types:

- Data types define bitwidth requirements
- HLS optimizes bitwidth except for function parameters

► Loops:

- Define iterative execution regions that can share HW resources

► Arrays:

- Main way of defining memory and data storage

► Operators:

- Implementations optimized for performance
- Automatically shared where possible to reduce area

```
void fir (
    data_t *y,
    coef_t c[4],
    data_t x
) {
    static data_t shift_reg[4];
    acc_t acc;
    int i;
    acc=0;
    loop: for (i=3;i>=0;i--) {
        if (i==0) {
            acc=x*c[0];
            shift_reg[0]=x;
        } else {
            shift_reg[i]=shift_reg[i-1];
            acc+=shift_reg[i] * c[i];
        }
    }
    *y=acc;
}
```

Control Defined by a Program

Code

```
void fir (
    data_t *y,
    coef_t c[4],
    data_t x
){
    static data_t shift_reg[4];
    acc_t acc;
    int i;

    acc=0;
    loop: for (i=3;i>=0;i--) {
        if (i==0) {
            acc+=x*c[0];
            shift_reg[0]=x;
        } else {
            shift_reg[i]=shift_reg[i-1];
            acc+=shift_reg[i]*c[i];
        }
    }
    *y=acc;
}
```

Control Behavior

Finite State Machine (FSM) states



Function Start

For-Loop Start

For-Loop End

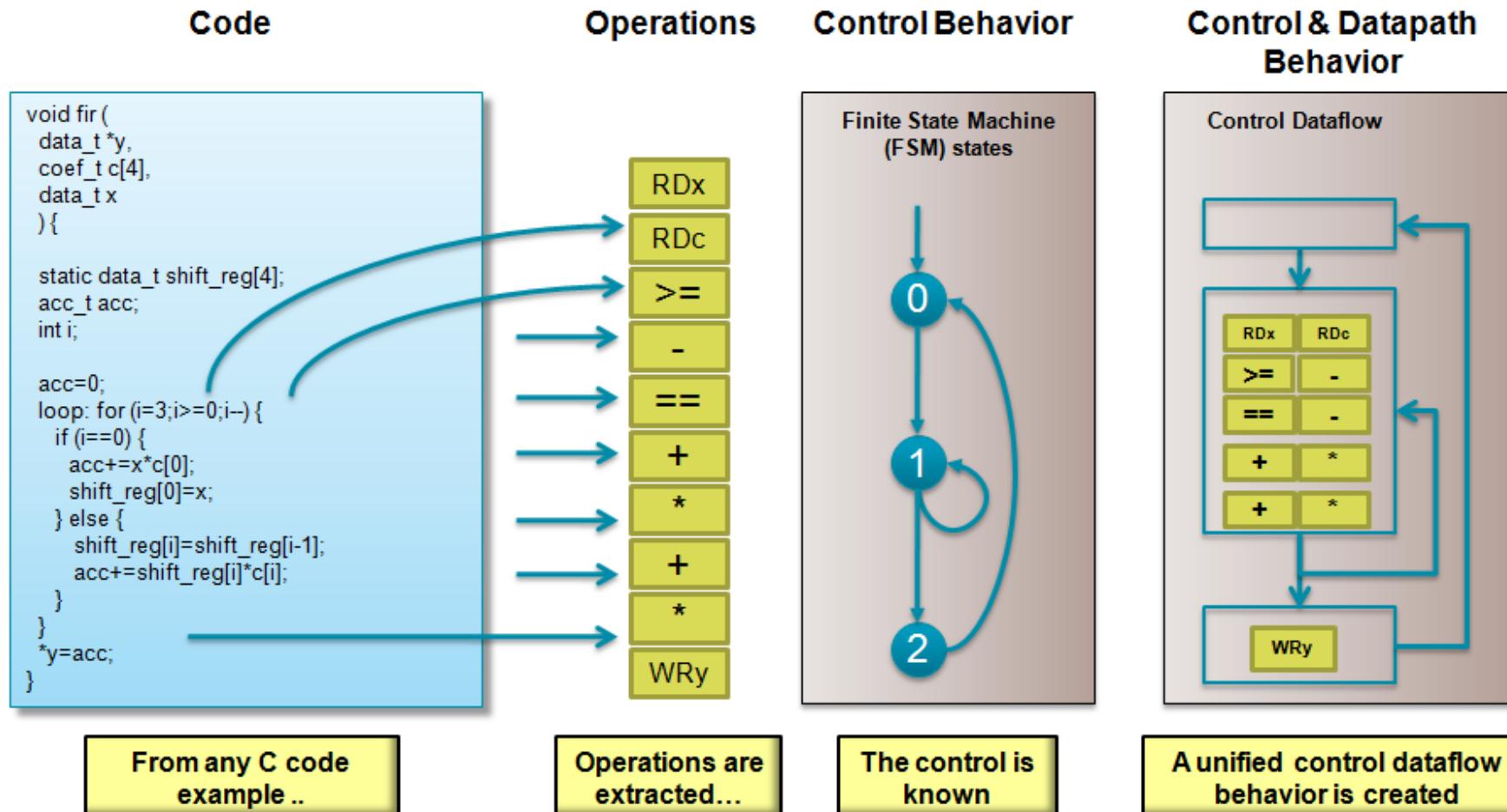
Function End

From any C code example ..

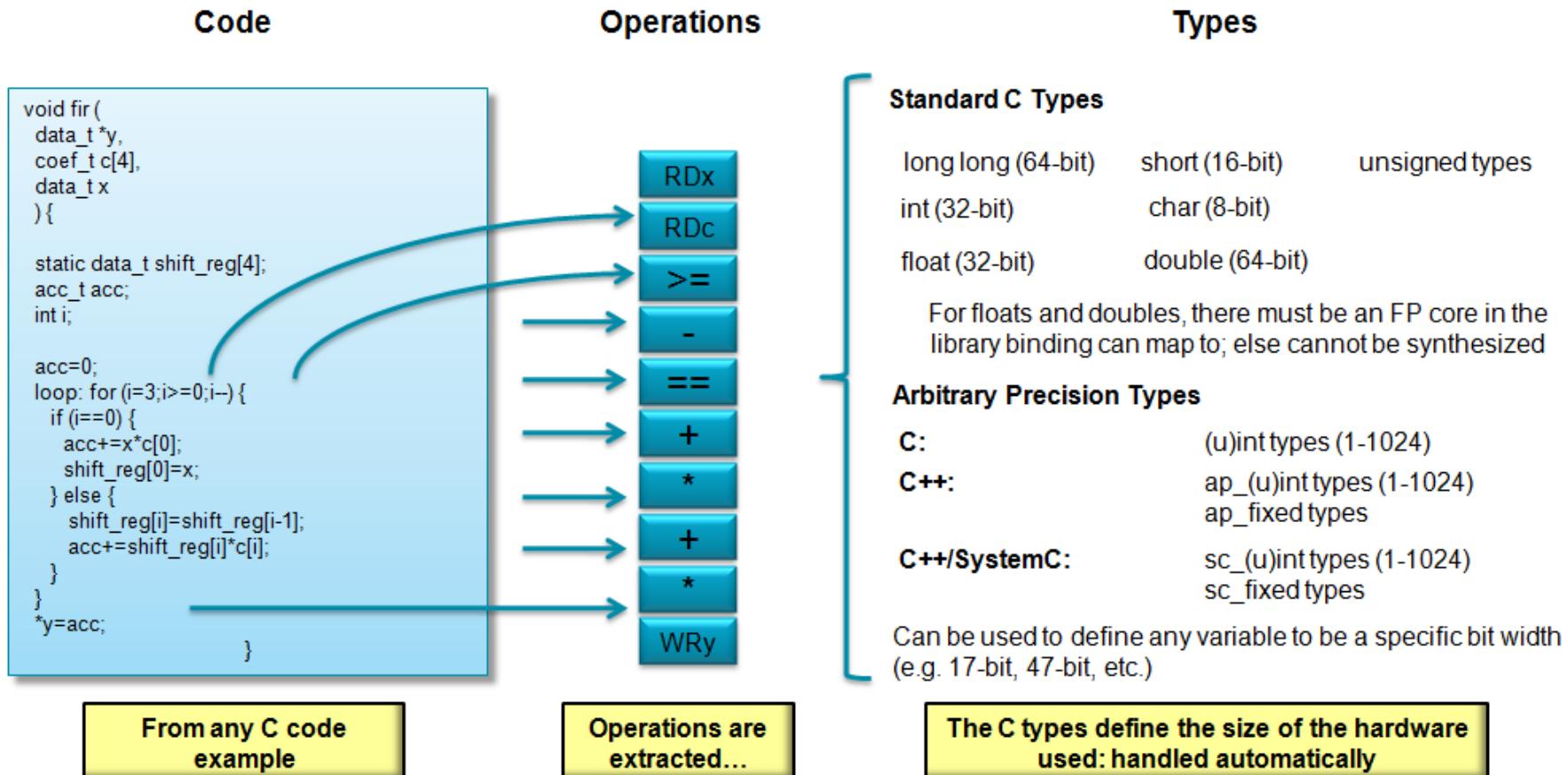
The loops in the C code correlated to states of behavior

This behavior is extracted into a hardware state machine

Combining Control and Operations

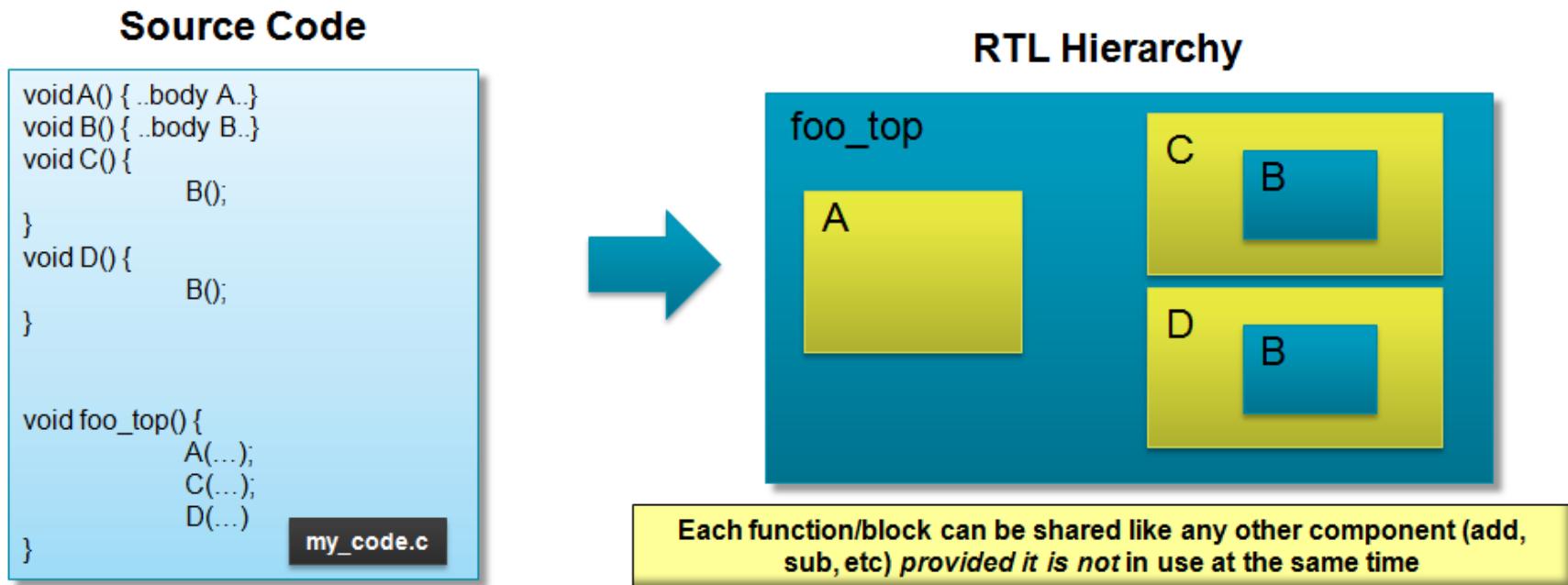


Optimizing and Sizing Program Operations



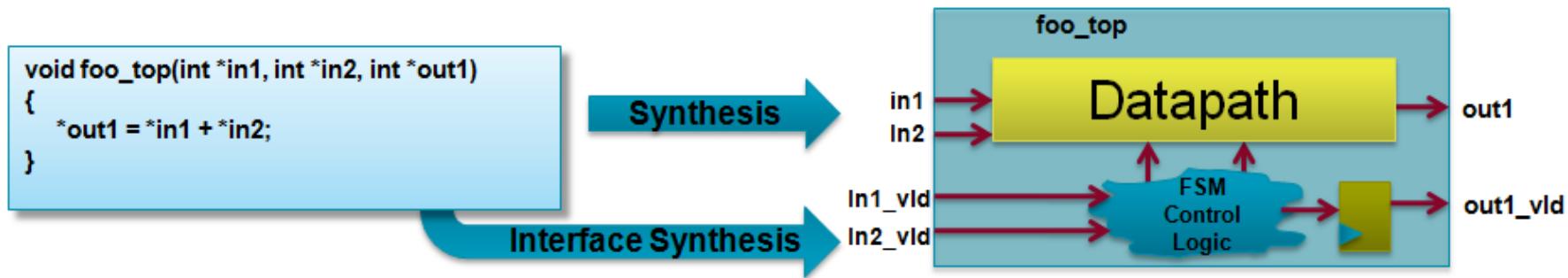
Program Functions and RTL Modules

- ▶ Functions are by default converted into RTL modules
- ▶ Functions define hierarchy in RTL
- ▶ Functions at the same hierarchical level can be shared like any operator to reduce resource consumption
 - Performance requirements control the level possible sharing



Completing a Design – I/O Port Creation

- Function parameters define data I/O ports and default protocols
 - Pointers → AXI4-Master interface
 - Scalars → AXI4-Lite interface or raw wires
 - Arrays → AXI4-Lite or AXI4 stream interface
- Protocol in generated HW are controlled through user directives





HLS Coding and Design Capture

Vivado HLS Development Environment

► A complete C validation and verification environment

- Vivado HLS supports complete bit-accurate validation of the C model
- Vivado HLS provides a productive C-RTL co-simulation verification solution

► Vivado HLS supports C, C++, and SystemC

- Functions can be written in any version of C
- Wide support for coding constructs in all three variants of C

► Modeling with bit accuracy

- Supports arbitrary precision types for all input languages
- Allows the exact bit widths to be modeled and synthesized

► Floating-point support

- Support for the use of float and double in the code

► Pointers and streaming-based applications

Vivado HLS Tiered Verification Flow

➤ Two steps to verifying the design

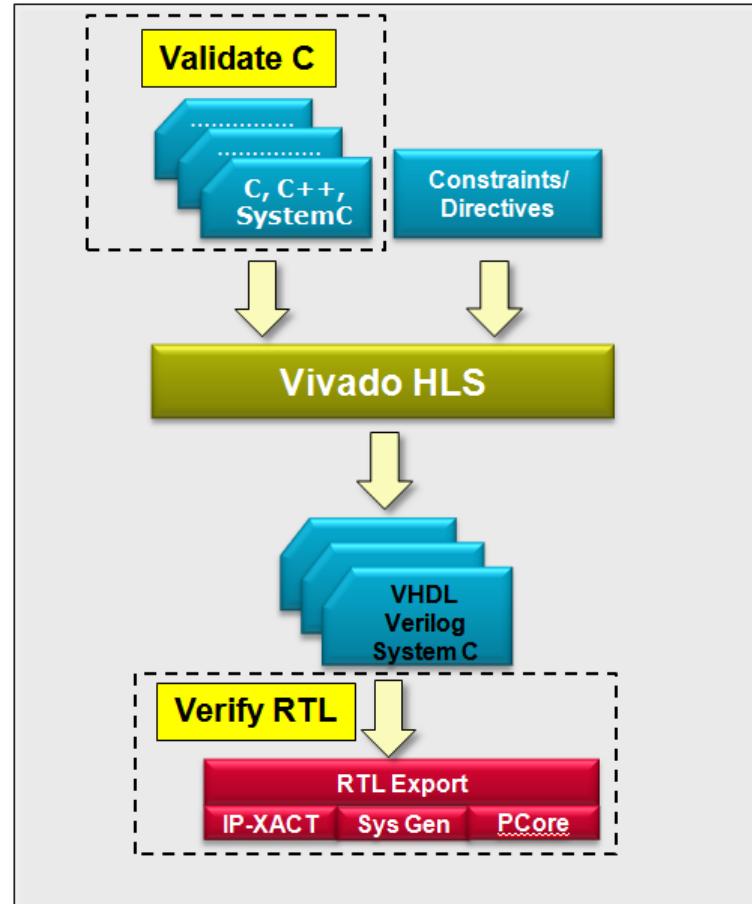
- Pre-synthesis: C validation
- Post-synthesis: RTL verification

➤ C validation

- Fast and free verification on any Operating System
- Prove algorithm correctness before RTL generation

➤ RTL Verification

- RTL Co-Simulation against the original program testbench



Coding Restrictions

➤ Data Types

- Forward declared data types
- Recursive type definitions

➤ Pointers

- General casting between user defined data types
- Pointers to dynamically allocated memory regions

➤ System Calls

- Dynamic memory allocation – must be replaced with static allocation
- Standard I/O and file I/O – automatically ignored by the compiler
- System calls
 - i.e. time(), sleep()

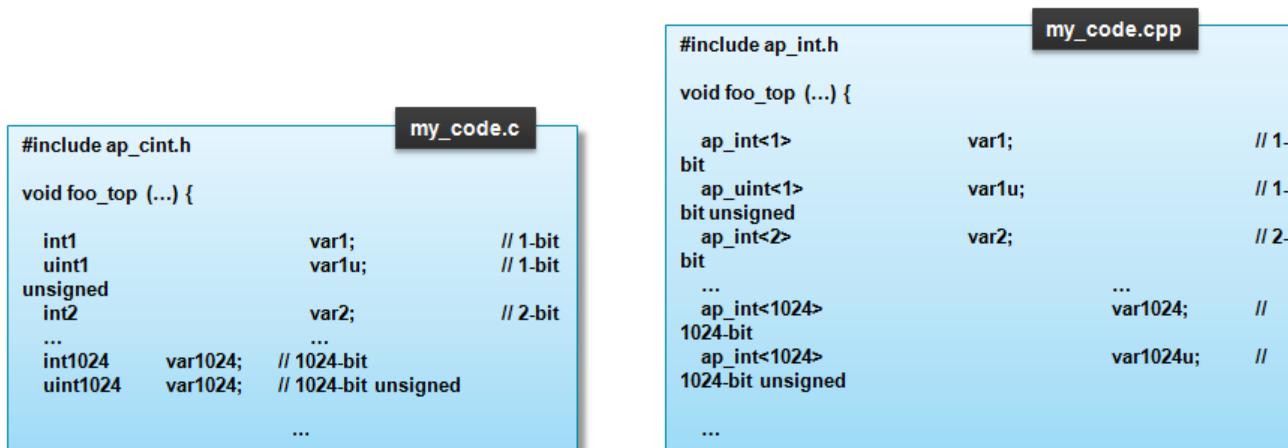
➤ Recursive functions that are not compile time bounded

➤ STL lib calls

- Not supported due to dynamic memory allocation
- Have compile time unbounded recursion

Arbitrary Precision Types

- C and C++ standard types are supported but limit the hardware
 - 8-bit, 16-bit, 32-bit boundaries
- Real hardware implementations use a wide range of bitwidths
 - Tailored to reduce hardware resources
 - Minimum precision to keep algorithm correctness
- HLS provides bit-accurate types in C and C++
 - SystemC and HLS types supported to simulate hardware datapaths in C/C++

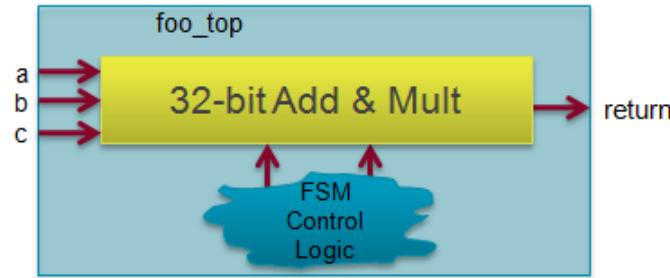


Algorithm Modeling with Arbitrary Types

► Code using native C types

```
int foo_top(int a, int b, int c)
{
    int sum, mult;
    sum=a+b;
    mult=sum*c;
    return mult;
}
```

Synthesis →

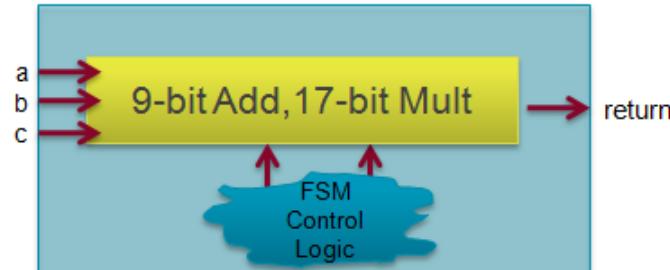


► Code using HLS types

- Software model matches hardware implementation
- C++ types can be compiled with both gcc and Visual Studio

```
int17 foo_top(int8 a, int8 b, int8 c)
{
    int9 sum;
    int17 mult;
    sum=a+b;
    mult=sum*c;
    return mult;
}
```

Synthesis →



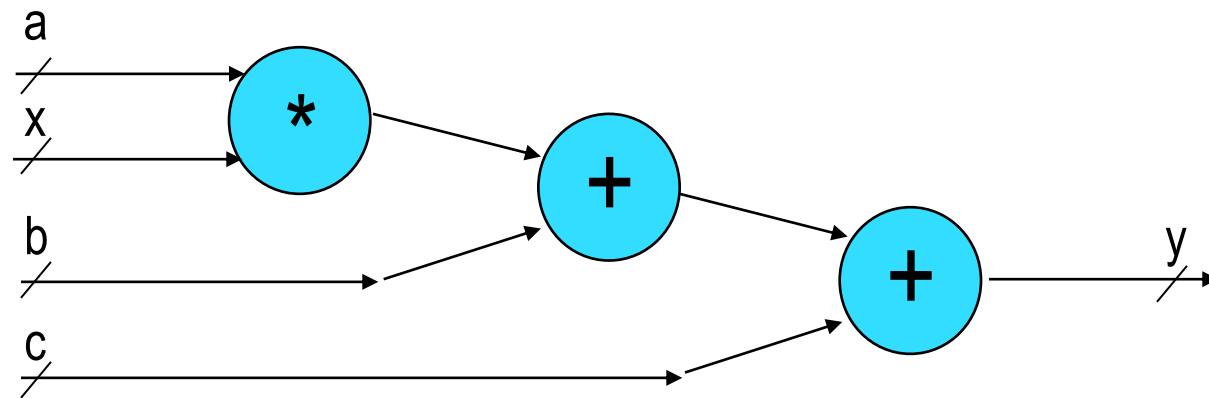


Default Behavior

Datapath Synthesis

- HLS begins by extracting a functional model of a C expression

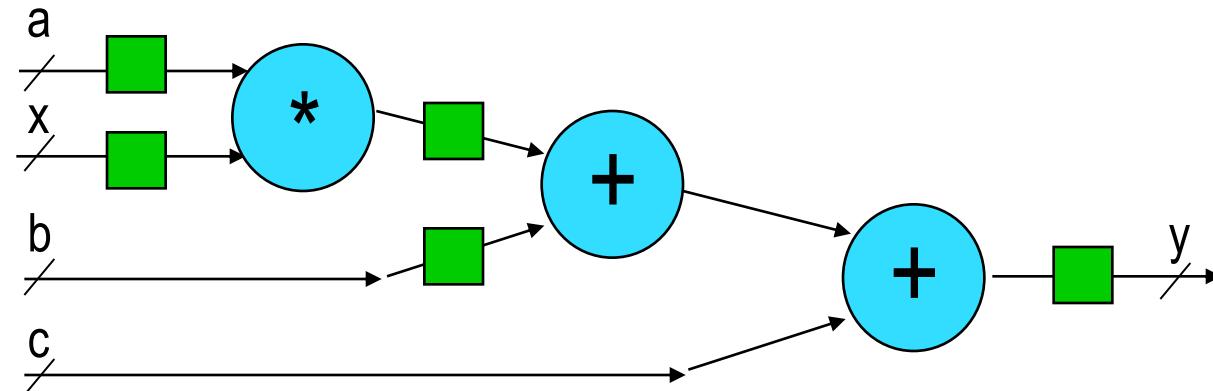
```
int a,b,c,x;  
int y = a*x + b + c
```



Datapath Synthesis - Pipelining

- HLS accounts for target frequency and device characteristics to determine minimum required pipelining
- Circuit will close timing but is not yet the optimal implementation

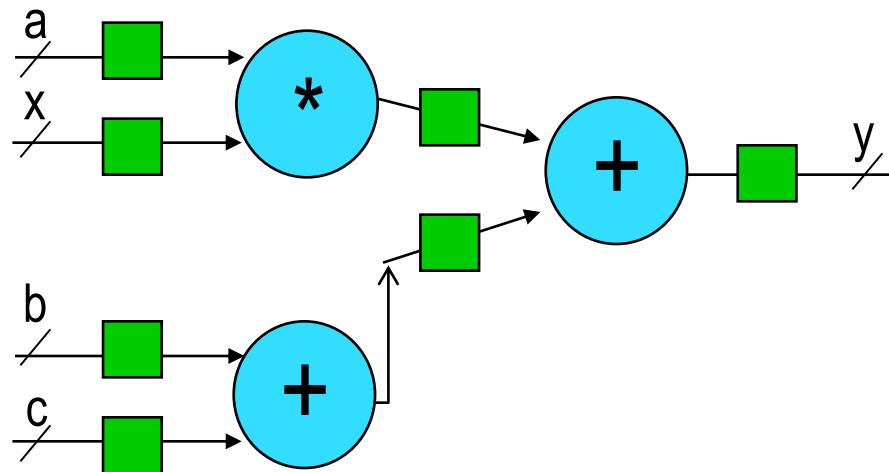
```
int a,b,c,x;  
int y = a*x + b + c
```



Datapath Synthesis - Optimization

- Automatic expression balancing for latency reduction
- Automatic restructuring to optimize use of FPGA fabric resources

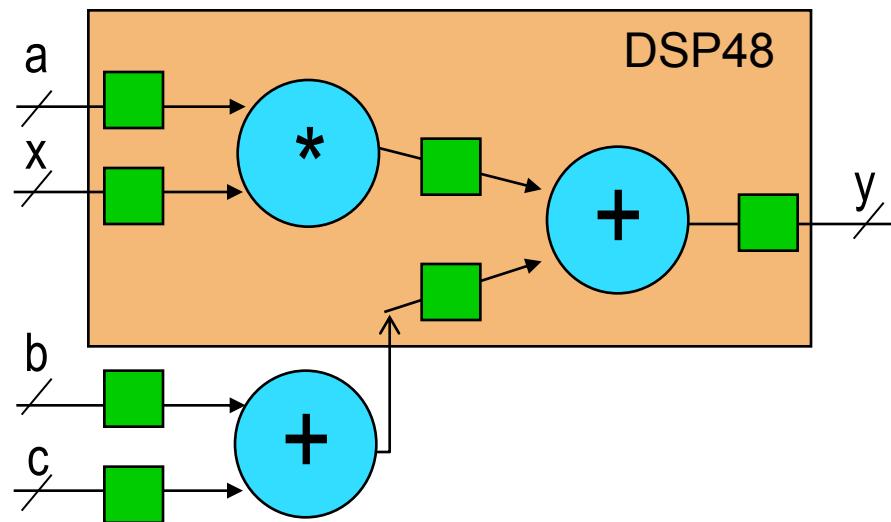
```
int a,b,c,x;  
int y = a*x + b + c
```



Datapath Synthesis – Predictable Implementation

- Restructuring from previous stage leads to optimized implementations using DSP48

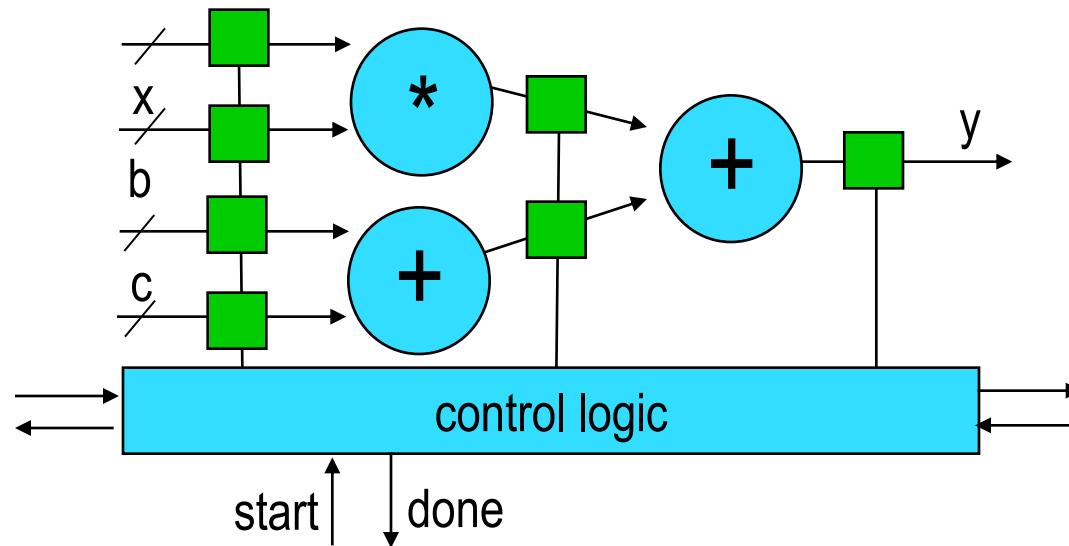
```
int a,b,c,x;  
int y = a*x + b + c
```



Datapath and Loops

- After a datapath is generated, loop control logic is added

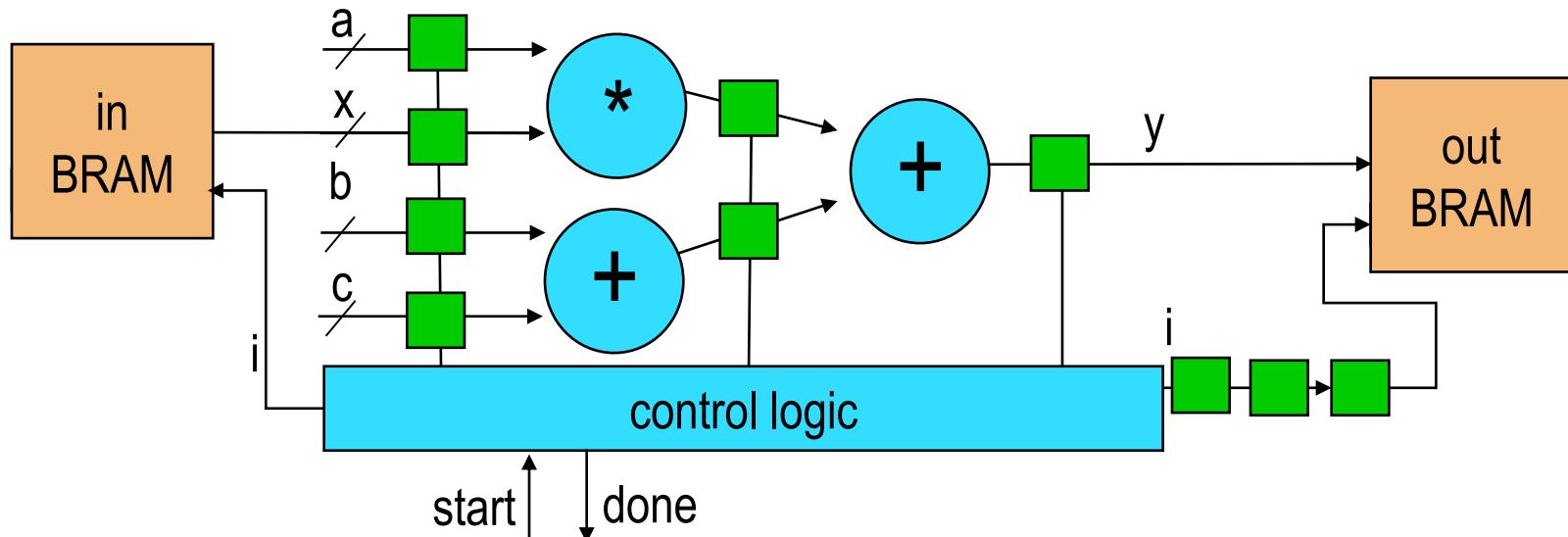
```
int a,b,c,x,y;  
for(int i = 0; i < 20; i++) {  
    x = get(); y = a*x + b + c; send(y);  
}
```



Defining Loop I/O with Arrays

- C arrays can be implemented as BRAMs or LUT-RAMs
- Default implementation depends on the depth and bitsize of the original C array

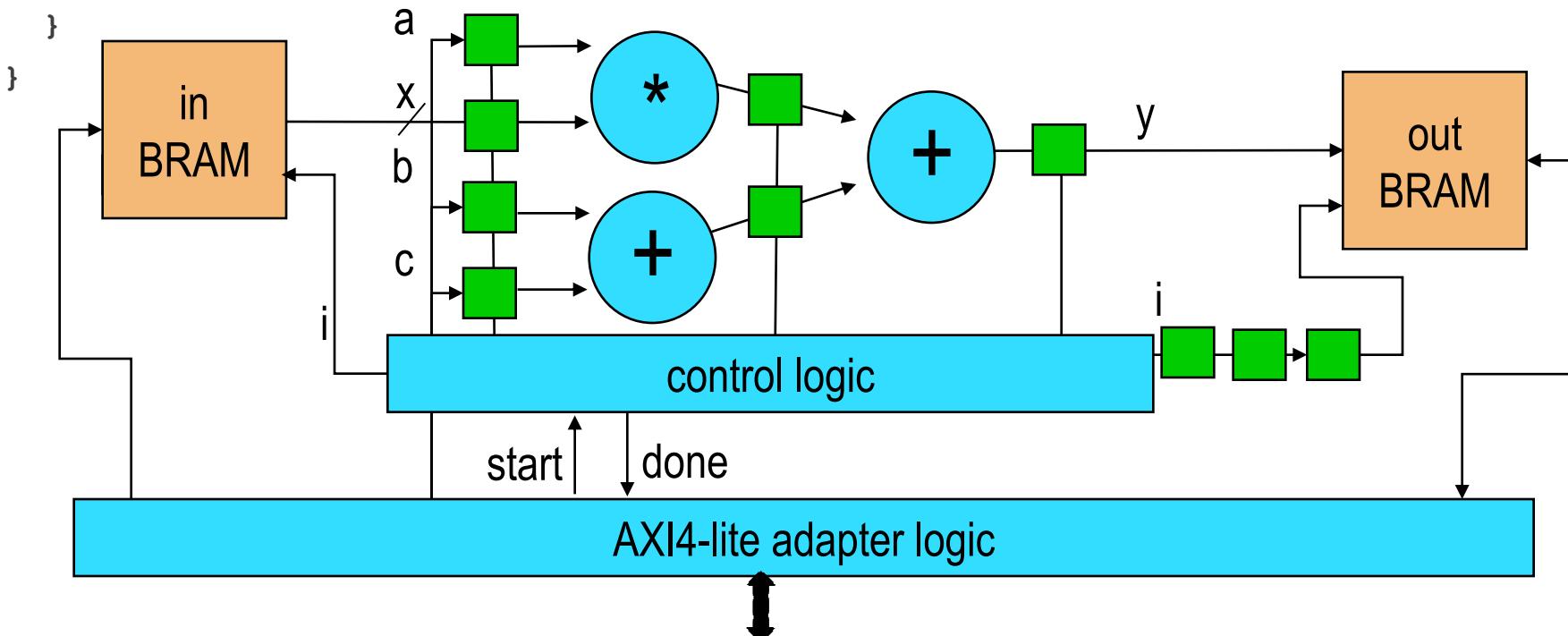
```
int a,b,c,x,y;  
for(int i = 0; i < 20; i++) {  
    x = in[i]; y = a*x + b + c; out[i] = y;  
}
```



Completing the Design – Interface Synthesis

- Function parameters become system level interfaces after HLS synthesis

```
f(int in[20], int out[20]) {  
    int a,b,c,x,y;  
    for(int i = 0; i < 20; i++) {  
        x = in[i]; y = a*x + b + c; out[i] = y;  
    }  
}
```





Performance Optimization

Arrays and Pointers

Arrays and Memory Bottlenecks

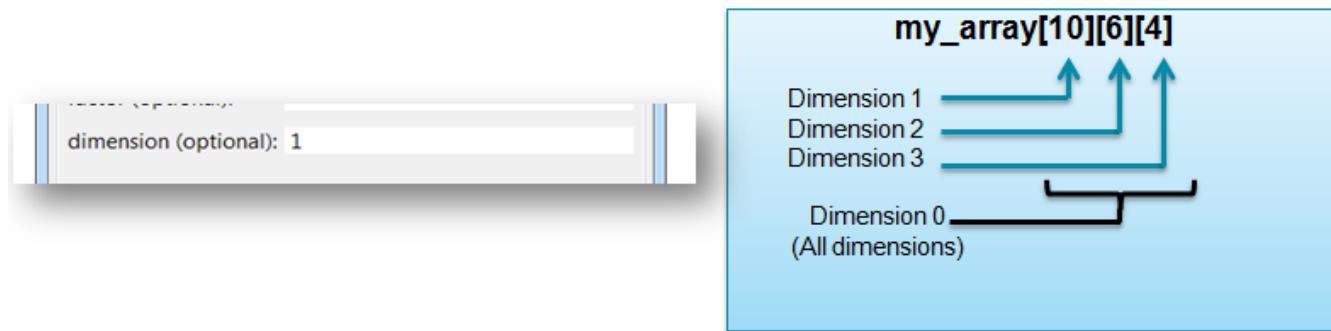
- Arrays are the basic construct to express memory to HLS
- Default number of memory ports defined by
 - Number of usages in the algorithm
 - Target throughput
- HLS default memory model assumes 2-port BRAMs
- Arrays can be reshaped and partitioned to remove bottlenecks
 - Changes to array layout do not require changes to the original code

```
void foo_top (...) {  
    ...  
    for (i = 2; i < N; i++)  
        mem[i] = mem[i-1] +mem[i-2];  
    }  
}
```



Even with a dual-port RAM, all reads and writes cannot be performed in one cycle

Array Optimization - Dimensions



► Examples: C array and RTL implementation

my_array[10][6][4] → partition dimension 3 →

my_array_0[10][6]
my_array_1[10][6]
my_array_2[10][6]
my_array_3[10][6]

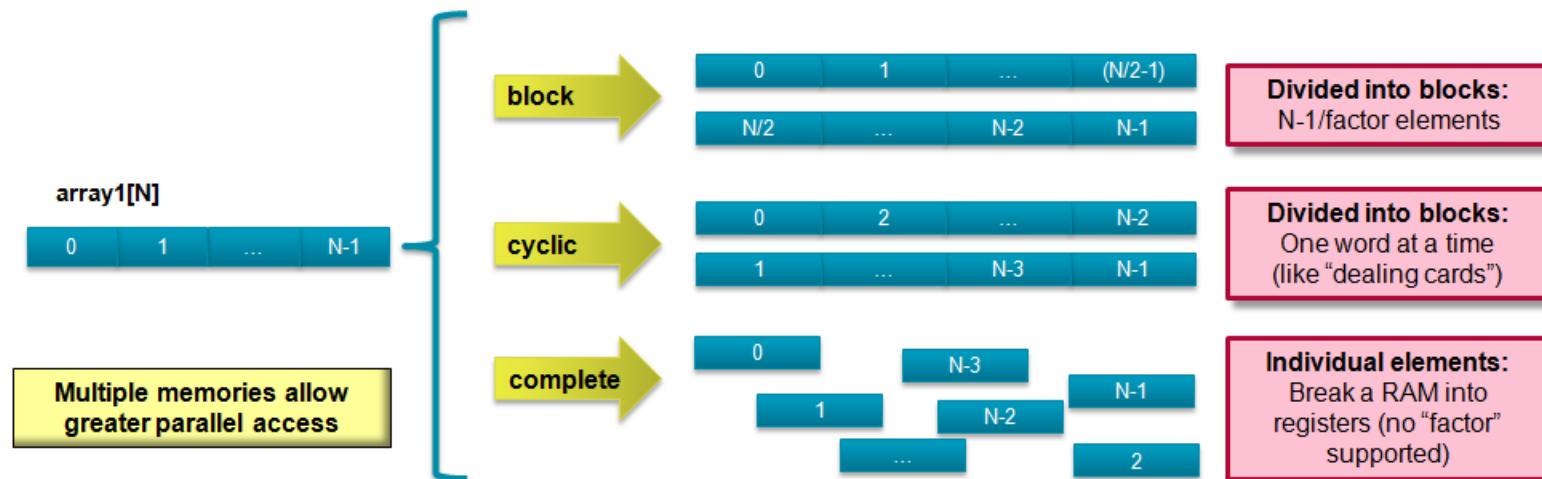
my_array[10][6][4] → partition dimension 1 →

my_array_0[6][4]
my_array_1[6][4]
my_array_2[6][4]
my_array_3[6][4]
my_array_4[6][4]
my_array_5[6][4]
my_array_6[6][4]
my_array_7[6][4]
my_array_8[6][4]
my_array_9[6][4]

my_array[10][6][4] → partition dimension 0 → $10 \times 6 \times 4 = 240$ individual registers

Array Optimization - Partitioning

- Partitioning splits arrays into independent memory banks in RTL
- Arrays can be partitioned on any dimension
 - Multi-dimension arrays can be partitioned multiple times
 - Dimension 0 applies a partitioning command to all dimensions

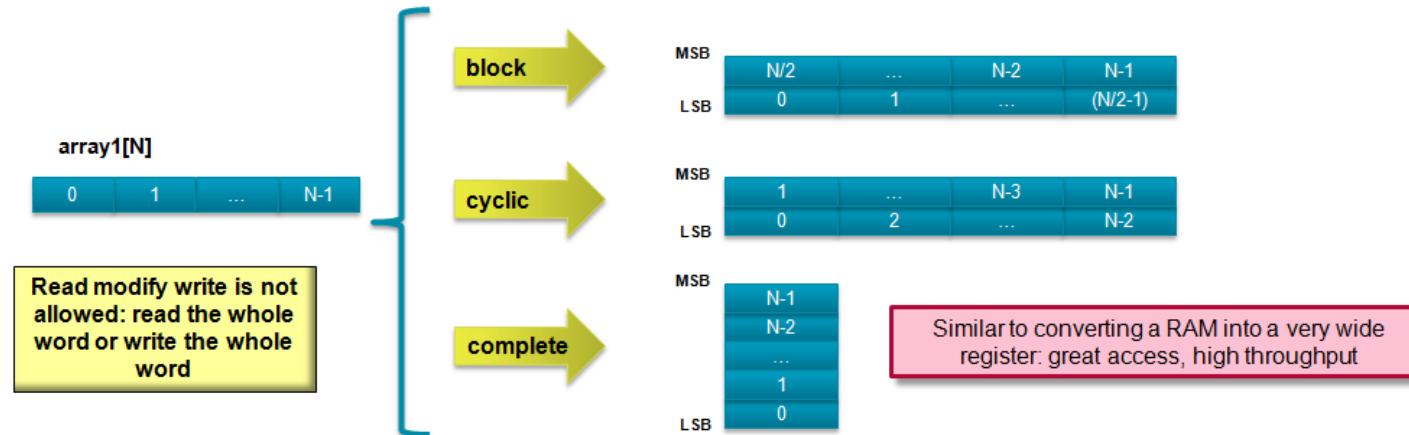


Array Optimization - Reshaping

► Reshaping combines

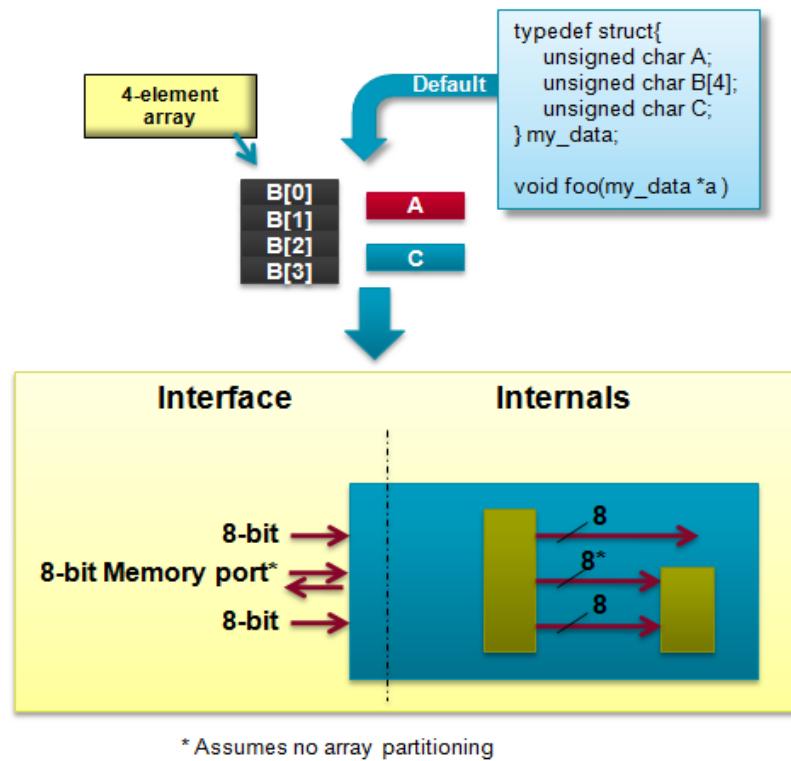
- Array entries into wider bitwidth containers
- Different arrays into a single physical memory

► New RTL level memories are automatically generated without changes to the C algorithm



Array Optimization – Structs

- ▶ Structs can contain any mix of arrays and scalar values
- ▶ Structs are automatically partitioned into individual elements
 - Each struct variable becomes a separate port or data bus
 - Independent control logic for each struct member

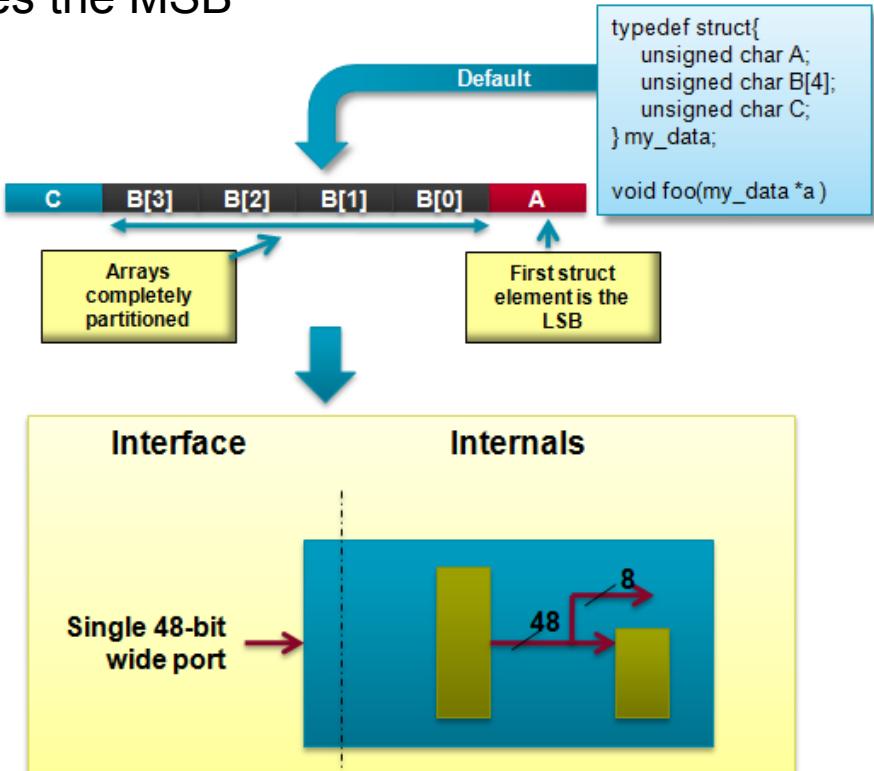


Array Optimization – Structs and Data Packing

► Data packing creates a single wide bus for all struct members

Bus Structure

- First element in the struct becomes the LSB
- Last element in the struct becomes the MSB
- Arrays are partitioned completely



Array Optimization - Initialization

► Example array initialization



- Implies coeff is initialized at the start of each function call
- Every function call has an overhead in writing the contents of the coeff BRAM

► Using static keyword moves initialization to bitstream

- Values of coeff are part of the FPGA configuration bitstream
- No function initialization overhead



Pointer Optimization – Access Mode

► Standard mode

- Each access results in a bus transaction
- Read and write operations can be mapped into a single transaction

Single read and write in standard mode

```
void foo (int *d) {  
    static int acc = 0;  
    int i;  
  
    acc += d[i];  
    d[i] = acc;  
}
```

Multiple reads and writes in standard mode

```
#include "bus.h"  
  
void foo (int *d) {  
    static int acc = 0;  
    int i;  
  
    for (i=0;i<4;i++) {  
        acc += d[i];  
        d[i] = acc;  
    }  
}
```

```
#include "bus.h"  
  
void foo (int *d) {  
    static int acc = 0;  
    int i;  
  
    for (i=0;i<4;i++) {  
        acc += *(d+i);  
        *(d+i) = acc;  
    }  
}
```

► Burst mode

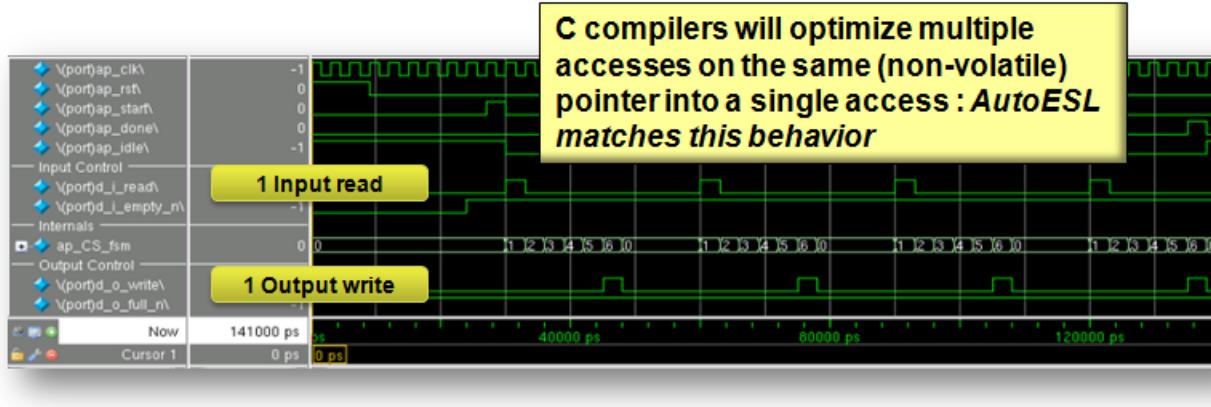
- Uses the C memcpy method
- Requires a local array inside the HLS block
 - Stores data for the burst write transaction
 - Stores data from the burst read transaction

Burst Mode

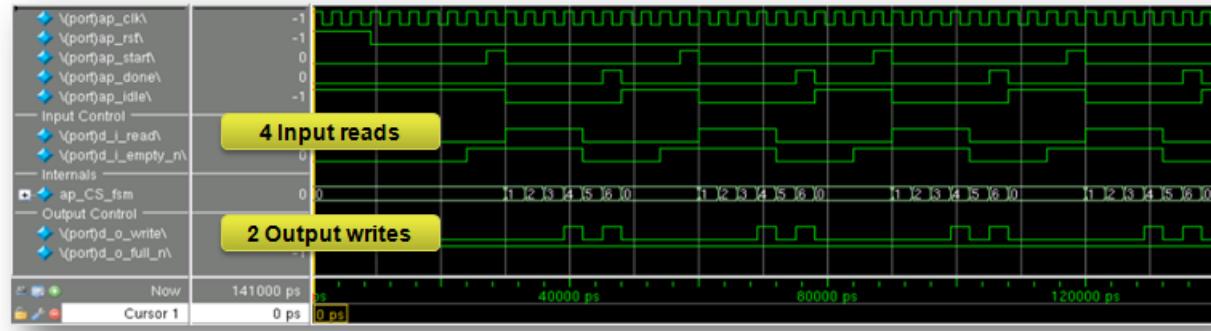
```
void foo (int *d) {  
    int buf1[4], buf2[4];  
    int i;  
  
    memcpy(buf1,d,4*sizeof(int));  
  
    for (i=0;i<4;i++) {  
        buf2[i] = buf1[3-i];  
    }  
  
    memcpy(d,buf2,4*sizeof(int));  
}
```

Pointer Optimization – Multiple Access

```
voidfifo(          int*d_o,  
              int*d_i  
{  
    static int acc = 0;  
    int cnt;  
  
    acc += *d_i;  
    acc += *d_i;  
    *d_o = acc;  
  
    acc += *d_i;  
    acc += *d_i;  
    *d_o = acc;  
}
```



```
voidfifo(          volatile int *d_o,  
              volatile int *d_i  
{  
    static int acc = 0;  
    int cnt;  
  
    acc += *d_i;  
    acc += *d_i;  
    *d_o = acc;  
  
    acc += *d_i;  
    acc += *d_i;  
    *d_o = acc;  
}
```





Performance Optimization

Loops

Loops - Classification

► Only perfect and semi-perfect loops are automatically optimized

► Perfect loops

- Computation expressed only in the inner most loop
- No initializations between loop statements
- Loop bounds are constant

► Semi-perfect loops

- Computation expressed only in the inner most loop
- No initializations between loop statements
- Loop bounds can be variable

► Other types of loops

- User needs to convert the loop into perfect or semi-perfect loop

```
Loop_outer: for (i=3;i>=0;i--) {  
    Loop_inner: for (j=3;j>=0;j--) {  
        [loop body]  
    }  
}
```

```
Loop_outer: for (i=3;i>N;i--) {  
    Loop_inner: for (j=3;j>=0;j--) {  
        [loop body]  
    }  
}
```

```
Loop_outer: for (i=3;i>N;i--) {  
    [loop body]   
    Loop_inner: for (j=3;j>=M;j--) {  
        [loop body]   
    }  
}
```

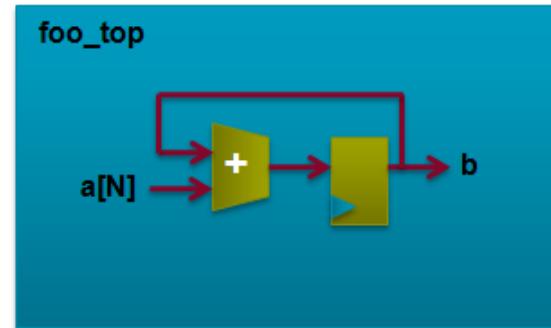
Loop Default Behavior

- Each loop iteration runs in the same HW state
- Each loop iteration runs on the same HW resources

```
void foo_top (...) {  
    ...  
    Add: for (i=3;i>=0;i--) {  
        b = a[i] + b;  
    }  
    ...  
}
```

Loops require labels if they are to be
referenced by Tcl directives
(GUI will auto-add labels)

Synthesis



Loops and Latency

- Loops enforce a minimum execution latency
- Incrementing the loop counter always consumes 1 clock cycle

```
void foo_top (...) {  
    ...  
    Add: for (i=3;i>=0;i--) {  
        b = a[i] + b;  
    ...  
}
```



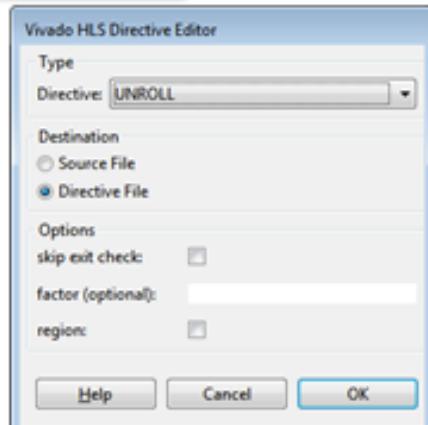
- Regardless of loop body, example will always take at least 4 clock cycles

Loops – Unrolling to Reduce Latency

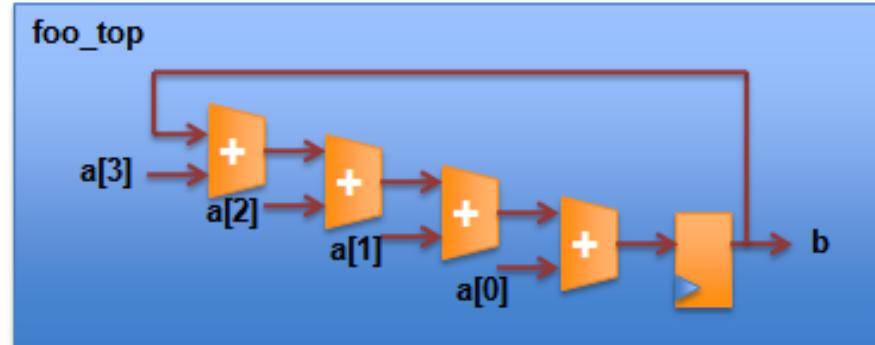
```
void foo_top (...) {  
    ...  
    Add: for (i=3;i>=0;i--) {  
        b = a[i] + b;  
    }  
    ...  
}
```

Select loop "Add" in the directives pane, right-click & Insert Directive then select unroll

Unrolled



Options explained
on next slide



clk

Option 1



Option 2



Option 3



Unrolled loops
allow greater
option &
exploration

Unrolled loops are likely to result in more hardware resources and higher area

Loops – Partial Unrolling

- HLS can unroll any loop by a factor
- Example shows unrolling by a factor of 2

- If N is not known, HLS inserts an exit check to maintain algorithm correctness
- If N is known and fully divisible by the unrolling factor
 - Exit check is removed

```
Add: for(int i = 0; i < N; i++) {  
    a[i] = b[i] + c[i];  
}
```

```
Add: for(int i = 0; i < N; i += 2) {  
    a[i] = b[i] + c[i];  
    if (i+1 >= N) break;  
    a[i+1] = b[i+1] + c[i+1];  
}
```

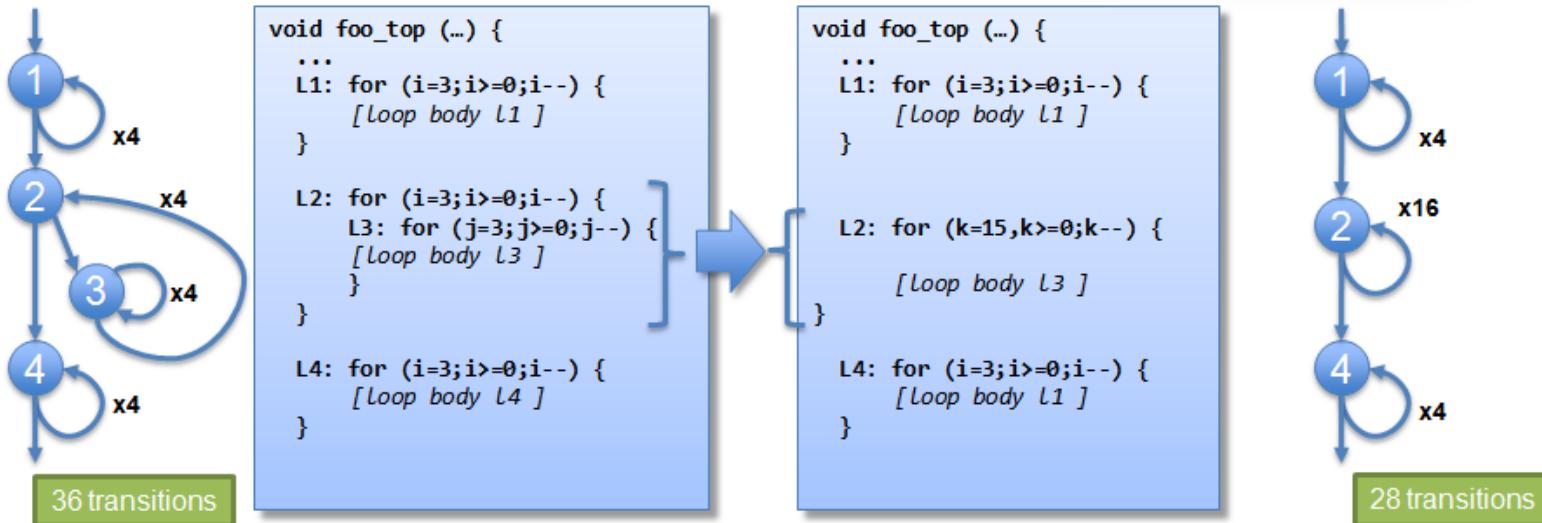
Effective code after compiler transformation

```
for(int i = 0; i < N; i += 2) {  
    a[i] = b[i] + c[i];  
    a[i+1] = b[i+1] + c[i+1];  
}
```

An extra adder for $N/2$ cycles trade-off

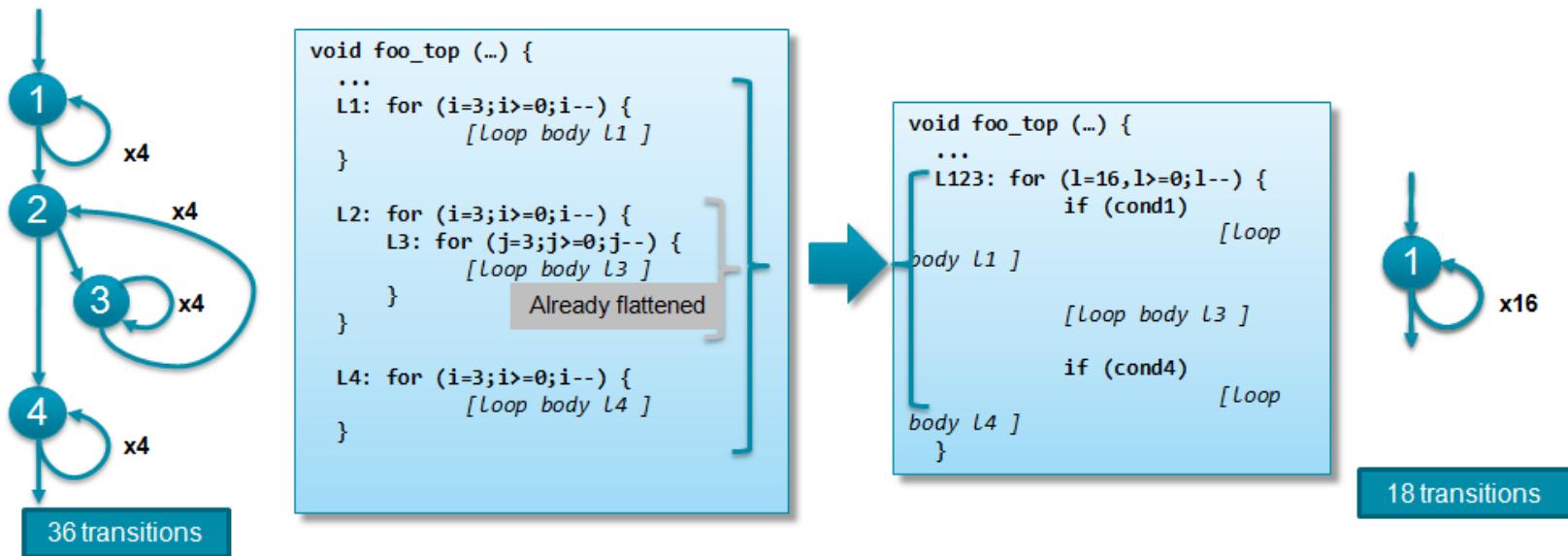
Loops - Flattening

- ▶ Perfect and semi-perfect loops are automatically flattened
 - Flattening eliminates state transitions between loop hierarchy levels
 - A loop state transition (counter increment) takes 1 clock cycle
- ▶ Automatic flattening can be turned off



Loops - Merging

- Loop merging reduces control regions in the generated RTL
- Does not require code changes as long as
 - All loops have either constant or variable bounds but not both
 - Loop body code always generates the same result regardless of how many times it is run
 - i.e $A = B$ is always the same - $A = A + 1$ depends on the loop iteration count



Loops – Merging Example

► Loop merging eliminates redundant computation

- Reduces latency
- Reduces resources

```
My_Region:{  
#pragma AP merge loop  
for (i = 0; i < N; ++i)  
    A[i] = B[i] + 1;  
for (i = 0; i < N; ++i)  
    C[i] = A[i] / 2;  
}
```

Merge

```
for (i = 0; i < N; ++i) {  
    A[i] = B[i] + 1;  
    C[i] = A[i] / 2;  
}
```

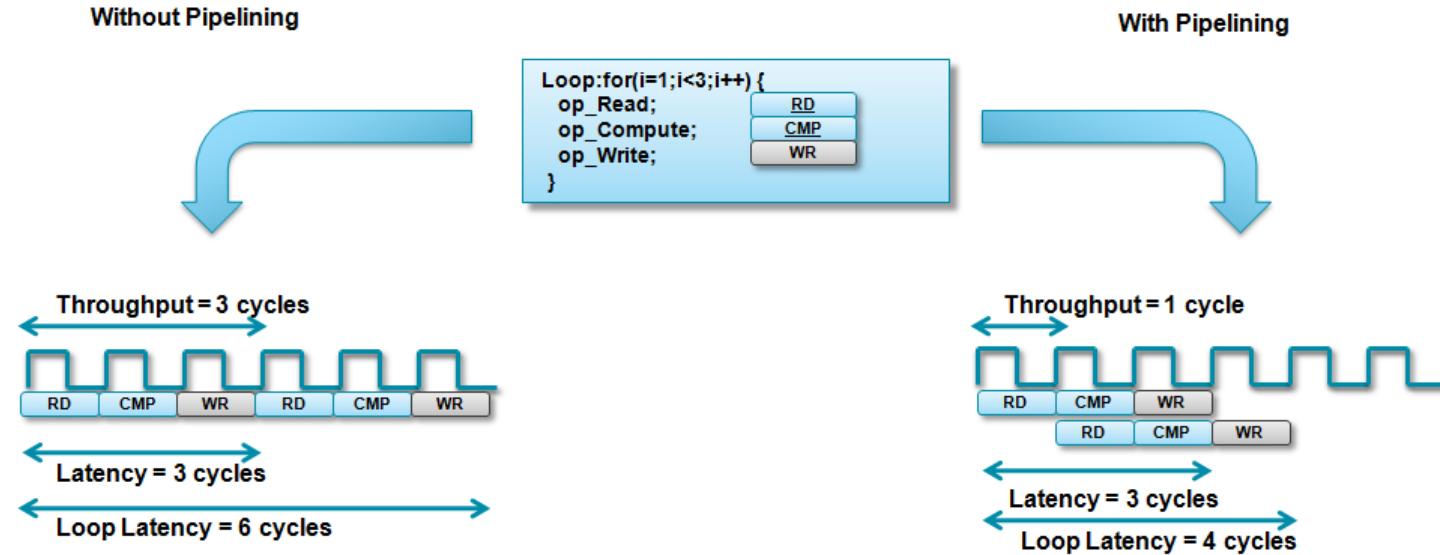
Effective code after compiler transformation

► Code implemented in RTL by HLS after merging

```
for (i = 0; i < N; ++i)  
    C[i] = (B[i] + 1) / 2;
```

Removes A[i], any address logic, and any potential memory accesses

Loops - Pipelining



- Loop iterations run sequentially
- Throughput = 3 clock cycles
- Latency
 - 3 cycles per iteration
 - 6 cycles for entire loop

- Loop iterations run in parallel
- Throughput = 1 clock cycle
- Latency
 - 3 cycles per iteration
 - 4 cycles for entire loop

Loops - Initiation Interval (II)

- The number of clock cycles between start of new loop body.



- II=1: one loop body per clock cycle
 - a ‘fully pipelined’ datapath for the loop body
- II=2: one loop body every 2 clock cycles
 - Allows for resource sharing of operators.

Loops – Hierarchy and II

► II is expressed by the PIPELINE directive

- Default value for PIPELINE = 1

► Can be applied to any level of a loop hierarchy

- Forces unrolling of any loop below the location of PIPELINE directive
- Increases parallelism and resources in a generated implementation
- Should be applied at a level that matches the input data rate of the design

```
void foo(in1[ ][], in2[ ][], ...){  
...  
L1:for(i=1;i<N;i++) {  
    L2:for(j=0;j<M;j++) {  
#pragma AP PIPELINE  
        out[i][j] = in1[i][j] + in2[i][j];  
    }  
}  
}
```

1adder, 3 accesses

```
void foo(in1[ ][], in2[ ][], ...){  
...  
L1:for(i=1;i<N;i++) {  
#pragma AP PIPELINE  
    L2:for(j=0;j<M;j++) {  
        out[i][j] = in1[i][j] + in2[i][j];  
    }  
}
```

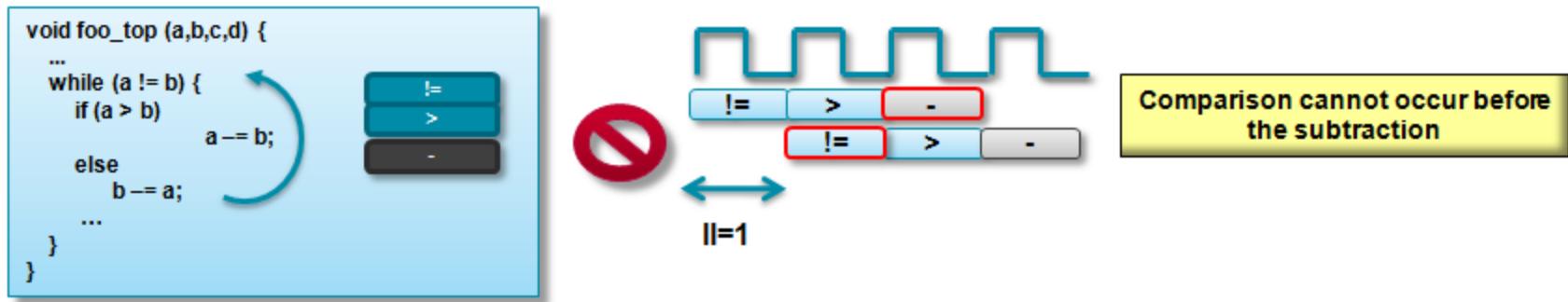
Unrolls L2
M adders, 3M accesses

```
void foo(in1[ ][], in2[ ][], ...){  
#pragma AP PIPELINE  
...  
L1:for(i=1;i<N;i++) {  
    L2:for(j=0;j<M;j++) {  
        out[i][j] = in1[i][j] + in2[i][j];  
    }  
}
```

Unrolls L1 and L2
N*M adders, 3(N*M) accesses

Loops – II and Feedback

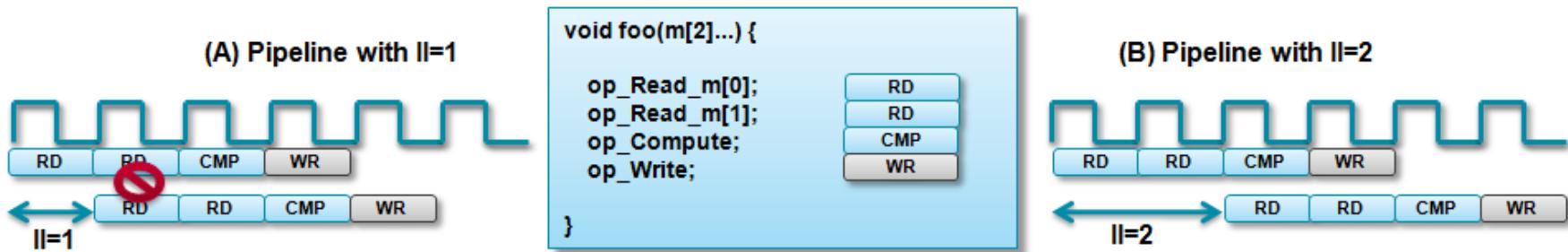
- Loop feedback is expressed as a dependence between iteration j to iteration j+1
- Type of dependence can limit pipelining
- If a dependence limits pipelining, HLS automatically relaxes the constraint



- User requested $\text{II} = 1$
- HLS generates $\text{II} = 2$ design due to dependence

Loops – II and Resource Contention

- HLS can instantiate all required resources to satisfy an II target within the boundaries of a generated module
- External ports can cause resource contention and are not automatically replicated
 - This type of contention can only be resolved by the user



- Memory `m` is a top level port
 - HLS assumes only 1 port is available to function `foo`
 - Multiple read operations push II from 1 to 2

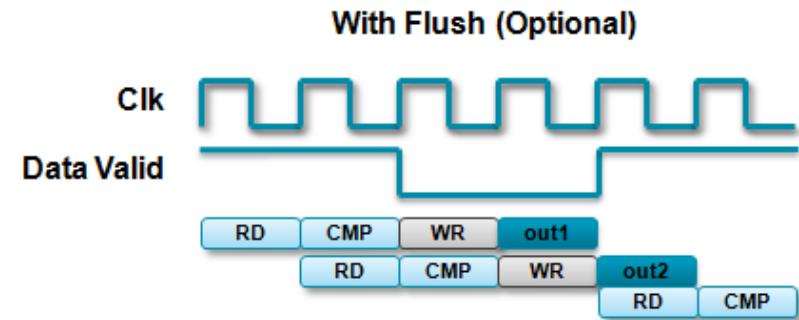
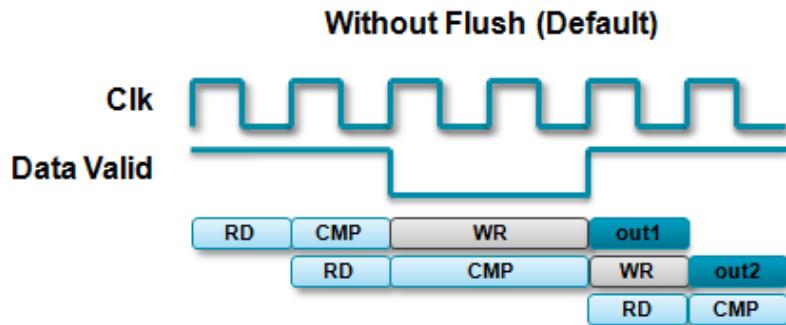
Loops – Pipeline Behavior

► HLS pipelines by default stall if the next input is not available

- For a loop, the next iteration doesn't start if the input data is not ready
- Stall affects all iterations currently being processed

► Default stall can be avoided with the flush option

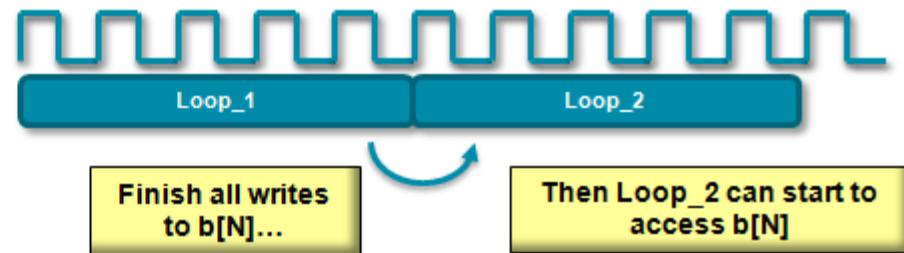
- Flushing the pipeline allows iterations to finish execution regardless of the state of the next iteration



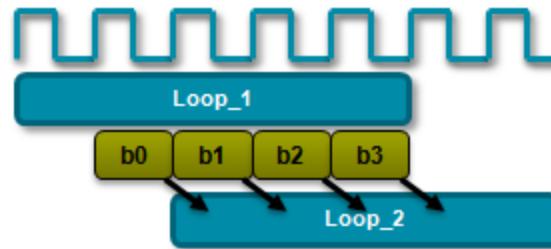
Loops - Dataflow

- Dataflow is the parallel execution of multiple loops within a function
- Loops to run in parallel communicate through arrays

```
int a[N], b[N], c[N];  
  
Loop_1: for (i=0;i<=N-1;i++) {      Loop_1  
    b[i] = a[i] + in1;  
}  
  
Loop_2: for (i=0;i<=N-1;i++) {      Loop_2  
    c[i] = b[i] * in2;  
}
```



- Arrays are changed to FIFOs to allow concurrent execution of Loop_1 and Loop_2





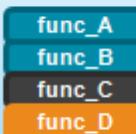
Performance Optimization

Functions

Functions

► For designs with multiple functions

```
void foo_top (a,b,c,d, *x, *y) {  
    ...  
    func_A(a,b,t1);  
    func_B(a,t1,t2);  
    func_C(c,t2,&x)  
    func_D(d,x,&y)  
}
```



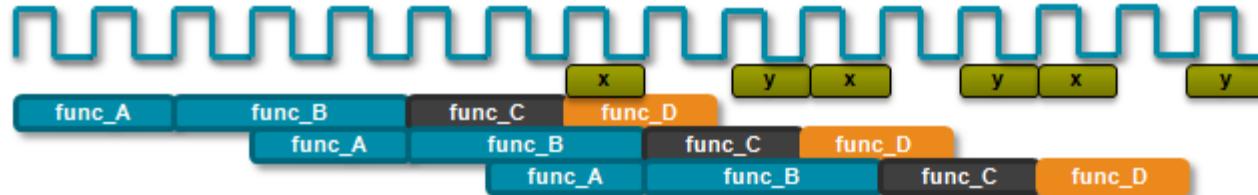
► Default Scheduling



The latency is 9 cycles

The throughput is also 9 cycles

► Functions can also be dataflowed like in the case of loops



The latency is still 9 cycles

The throughput is now 3



Area Optimization

Functions and RTL Hierarchy

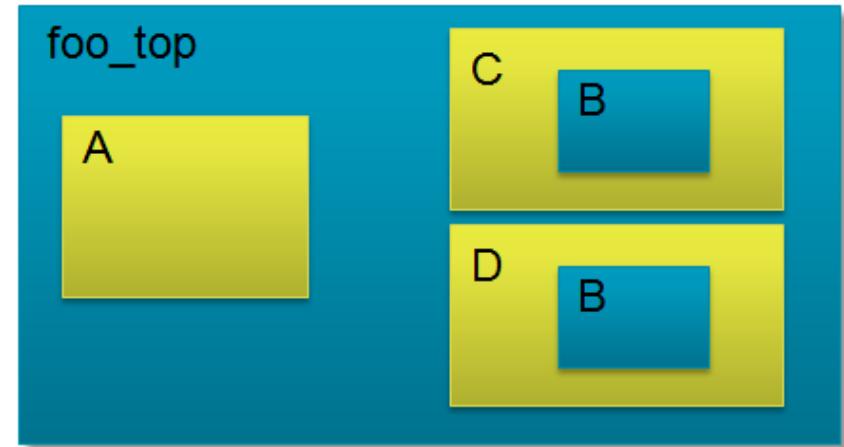
Source Code

```
void A() { ..body A..}
void B() { ..body B..}
void C() {
    B();
}
void D() {
    B();
}

void foo_top() {
    A(...);
    C(...);
    D(...);
}
```

my_code.c

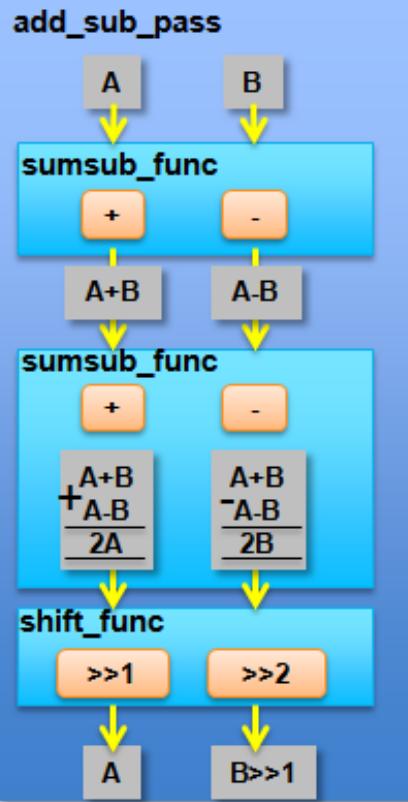
RTL Hierarchy



Functions can be inlined – the hierarchy was removed and the function dissolved into the surrounding function

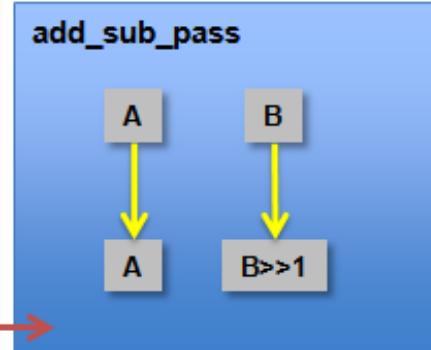
Function Inlining

No Inlining



```
int sumsub_func(int *in1, int *in2, int *outSum, int *outSub) {  
    *outSum = *in1 + *in2;  
    *outSub = *in1 - *in2;  
}  
  
int shift_func(int *in1, int *in2, int *outA, int *outB) {  
    *outA = *in1 >> 1;  
    *outB = *in2 >> 2;  
}  
  
void add_sub_pass(int A, int B, int *C, int *D) {  
    int apb, amb;  
    int a2, b2;  
  
    sumsub_func(&A,&B,&apb,&amb);  
    sumsub_func(&apb,&amb,&a2,&b2);  
    shift_func(&a2,&b2,C,D);  
}
```

Inlining



Zero Area

2 Adders
2 Subtractors

Inlining allows optimization to be performed across function hierarchies

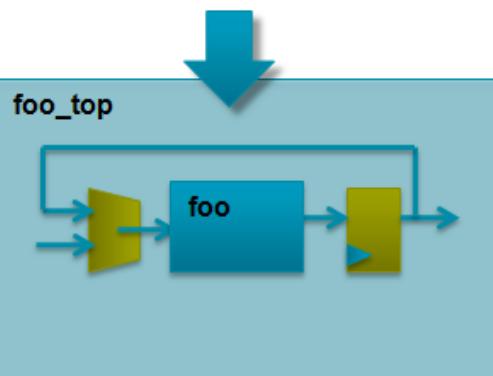
Like RTL ungrouping, too much inlining can create a lot of logic and slow runtime

Function Inlining and Allocation

Easy to Share

```
void foo() {  
}  
void foo_top() {  
    foo(...);  
    foo(...);  
}
```

```
set_directive_allocation -limit 1  
-type function foo_top foo
```

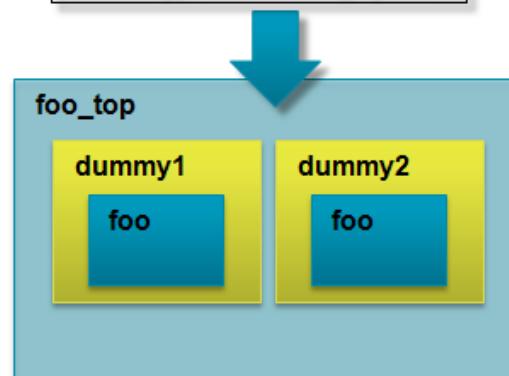


One RTL block is reused for both instances of function `foo`

Cannot be Shared

```
void dummy1() {  
    foo();  
}  
void dummy2() {  
    foo();  
}  
void foo_top() {  
    dummy1(...);  
    dummy2(...);  
}
```

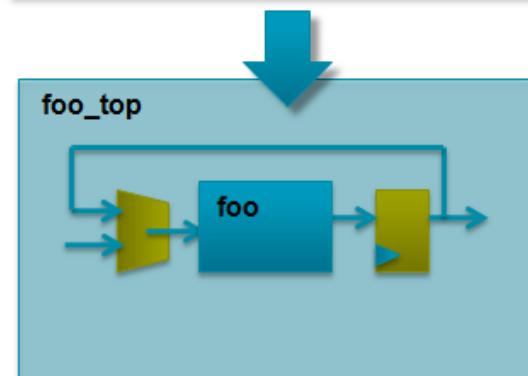
```
set_directive_allocation -limit 1  
-type function foo_top foo
```



Function `foo` is not within the immediate scope of `foo_top`

Controlling Sharing

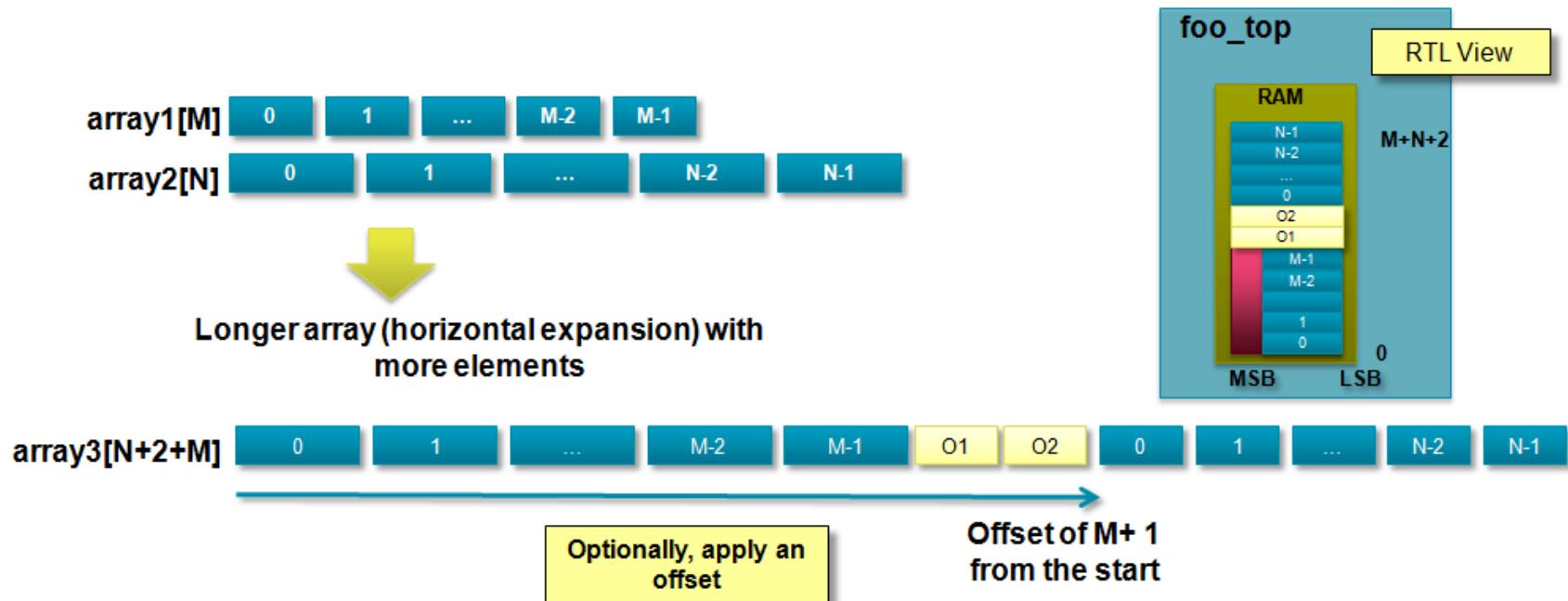
```
set_directive_allocation -limit 1  
-type function foo_top foo  
  
set_directive_inline dummy1  
set_directive_inline dummy2
```



Inlining brings `foo` into function `foo_top` where it can be shared

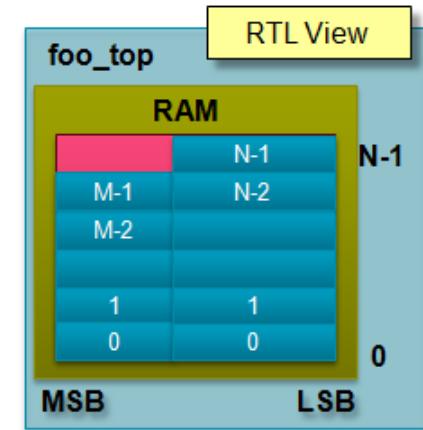
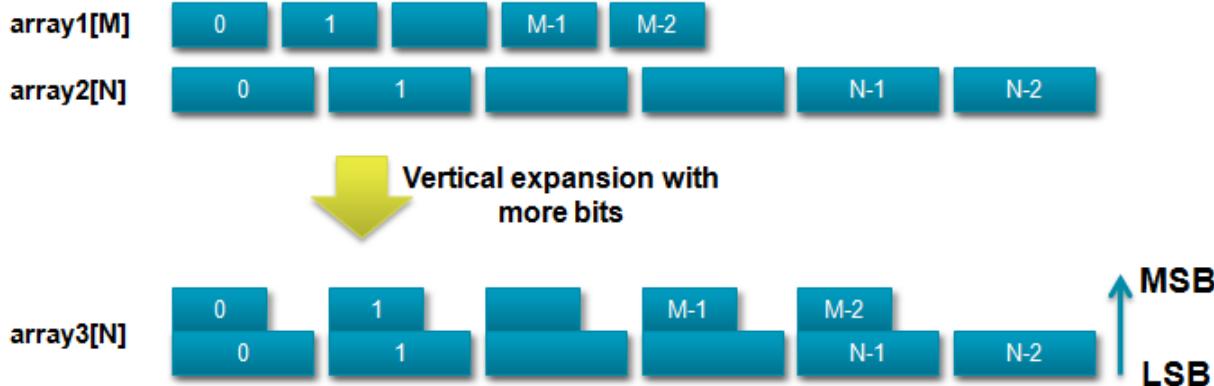
Array Mapping - Horizontal

- Combine multiple C arrays into 1 deeper memory
- Default is to concatenate arrays one after the other
 - User can introduce an offset to account for a system address map if the combined memories are top level ports



Array Mapping - Vertical

- Combine multiple C arrays into 1 wider memory
- Arrays use the same ordering as structs for packing
 - First array represents the LSB bits of the wider memory





Interface Definition

Default Ports

➤ Clock

- One clock per C/C++ design
- Multiple clocks possible for SystemC designs

➤ Reset

- Applies to FSM and variables initialized in the C algorithm

➤ Clock Enable

- Optional port
- One clock enable per design
- Attached to all modules within an HLS generated design

```
#include "adders.h"
int adders(int in1, int in2, int *sum) {
    int temp;
    *sum = in1 + in2 + *sum;
    temp = in1 + in2;
    return temp;
}
```

Synthesis



Function Parameters

➤ Function parameters

- Data ports for RTL I/O

➤ Function return

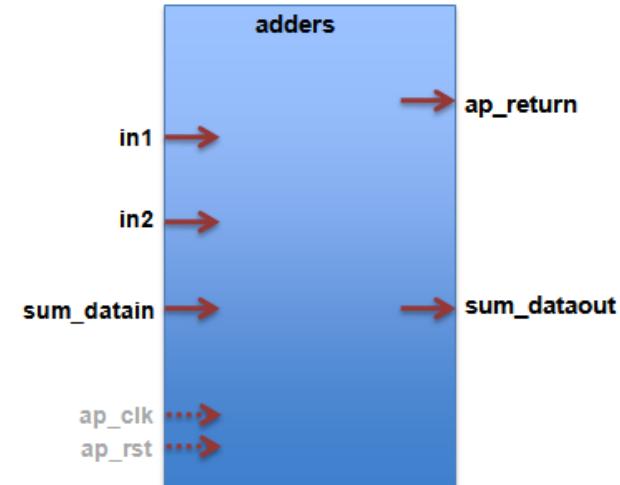
- 1 per HLS design
- Valid at the end of the C function call

➤ Pointers

- Can be implemented as both input and output
- Transformed into separate ports for each direction

```
#include "adders.h"
int adders(int in1, int in2, int *sum) {
    int temp;
    *sum = in1 + in2 + *sum;
    temp = in1 + in2;
    return temp;
}
```

Synthesis

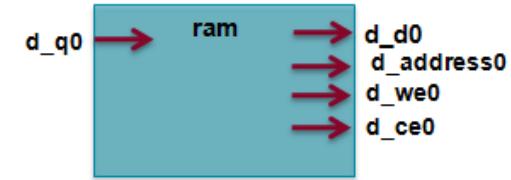


Function Parameters - Arrays

► RAM ports

- Default port for an array
- Assumes only 1 port connected to the HLS block
- Automatic generation of address and data ports

```
#include "ram.h"
void ram (int d[DEPTH], ...) {
    ...
}
```



► FIFO ports

- Example of streaming I/O
- Assumes array is accessed in sequential order

```
#include "fifo.h"
void fifo (int d_o[DEPTH],
           int d_i[DEPTH]) {
    ...
}
```



Block-Level Protocol

- By default all HLS generated designs have a master control interface

➤ ap_start

- Starts the RTL module, same as starting a function call in C

➤ ap_idle

- RTL module is idle

 ap_done

- RTL module completed a function call
 - The data in the ap_return port is valid

➤ ap_ready (not shown)

- Only generated for designs with top level function pipelining

Page 10 of A

[@newfunction](#)

 XILINX  ALL PROGRAMMABLE™

I/O Data Transfer Protocols

► Port I/O protocol

- Selected by the user to integrate the HLS generated block into a larger design
- Control the sequencing of data on a per interface basis

► Allows mapping to AXI and HLS provided protocols

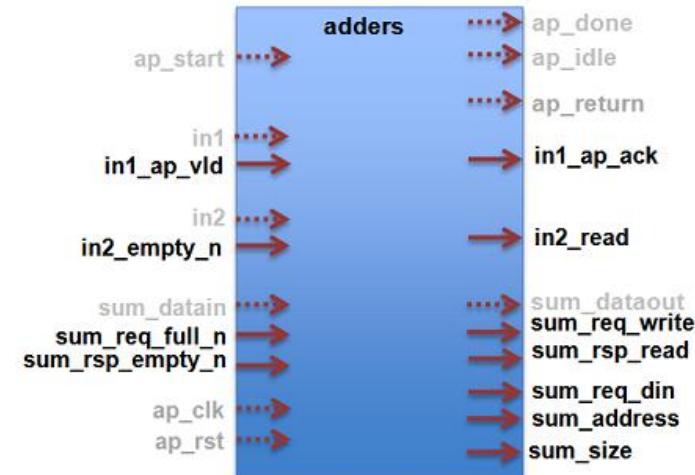
- Interface synthesis in C and C++ designs

► User can define their own interface protocol

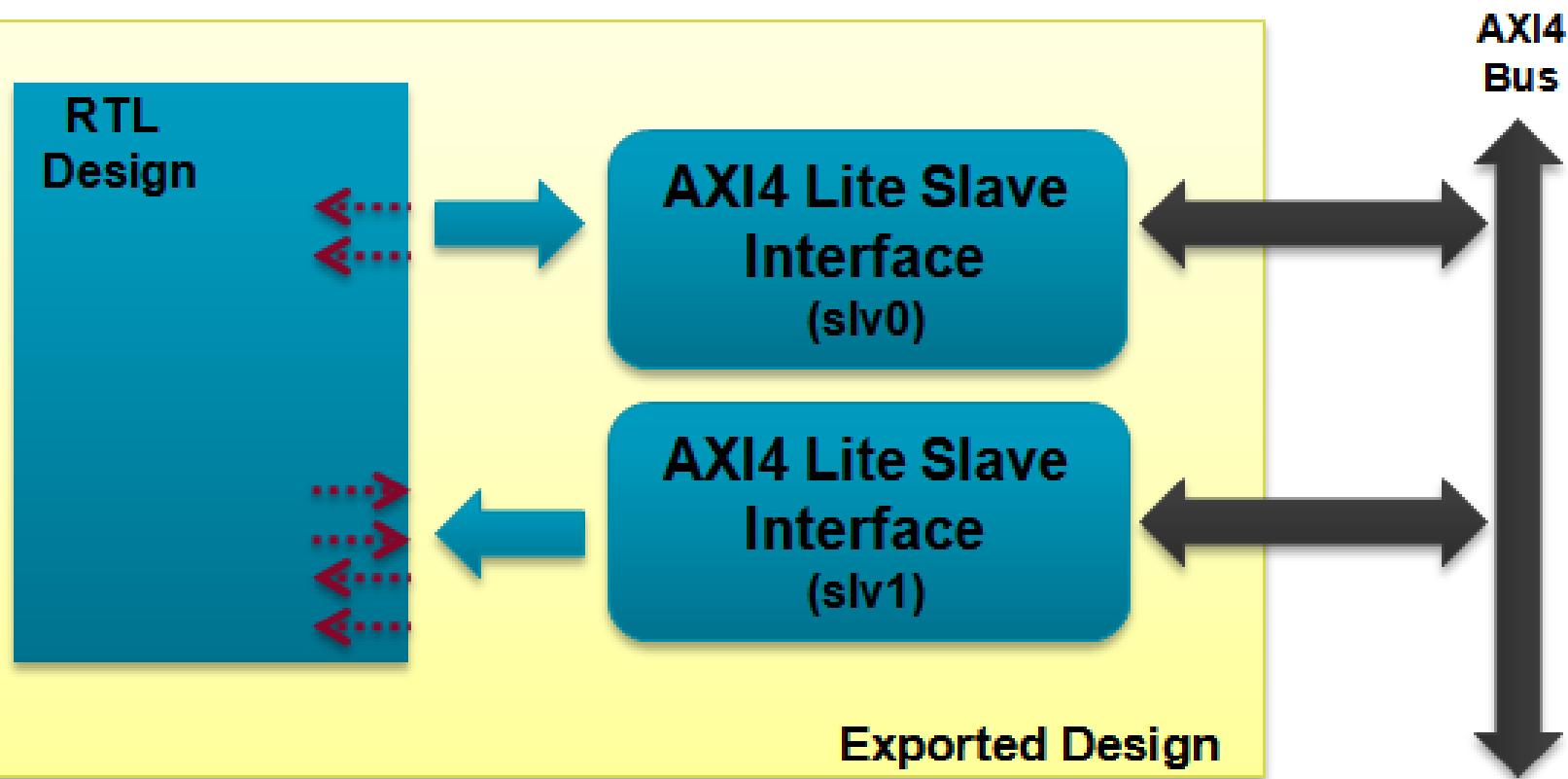
- SystemC designs natively express all port interfaces

```
#include "adders.h"
int adders(int in1, int in2, int *sum) {
    int temp;
    *sum = in1 + in2 + *sum;
    temp = in1 + in2;
    return temp;
}
```

Synthesis



Connecting to Standard Buses





Let's try it out!

16x16 Matrix Multiply

Basic AutoESL training in one slide

► Pick good places to pipeline.

- `#pragma HLS pipeline`

► Partition memories if needed.

- `#pragma HLS ARRAY_PARTITION variable=? complete dim=?`

► Watch for recurrences

- Might need to rewrite code or pick a different algorithm

► Use reduced-bitwidth operations.

- `ap_int<>, ap_uint<>, ap_fixed<>`

Basic Matrix Multiply

```
void mm(int in_a[A_ROWS] [A_COLS] ,
        int in_b[A_COLS] [B_COLS] ,
        int out_c[A_ROWS] [B_COLS])
{
    // matrix multiplication of a A*B matrix
    a_row_loop: for (int i = 0; i < A_ROWS; i++) {
        b_col_loop: for (int j = 0; j < B_COLS; j++) {
            int sum_mult = 0;
            a_col_loop: for (int k = 0; k < A_COLS; k++) {
                sum_mult += in_a[i][k] * in_b[k][j];
            }
            out_c[i][j] = sum_mult;
        }
    }
}
```

DSP48s: 3
Latency: 25121 clocks

Pipelined Matrix Multiply

```
void mm_pipelined(int in_a[A_ROWS] [A_COLS] ,
                  int in_b[A_COLS] [B_COLS] ,
                  int out_c[A_ROWS] [B_COLS])
{
    int sum_mult;

    // matrix multiplication of a A*B matrix
    a_row_loop: for (int i = 0; i < A_ROWS; i++) {
        b_col_loop: for (int j = 0; j < B_COLS; j++) {
            sum_mult = 0;
            a_col_loop: for (int k = 0; k < A_COLS; k++) {
                #pragma HLS pipeline
                sum_mult += in_a[i][k] * in_b[k][j];
            }
            out_c[i][j] = sum_mult;
        }
    }
}
```

DSP48s: 3
Latency: 6154 clocks
Loop II: 1

Parallel Dot-Product Matrix Multiply

```
void mm_parallel_dot_product(int in_a[A_ROWS][A_COLS],  
                            int in_b[A_COLS][B_COLS],  
                            int out_c[A_ROWS][B_COLS])  
{  
#pragma HLS ARRAY_PARTITION DIM=2 VARIABLE=in_a complete  
#pragma HLS ARRAY_PARTITION DIM=1 VARIABLE=in_b complete  
    int sum_mult;  
  
    // matrix multiplication of a A*B matrix  
    a_row_loop: for (int i = 0; i < A_ROWS; i++) {  
        b_col_loop: for (int j = 0; j < B_COLS; j++) {  
            #pragma HLS pipeline  
            sum_mult = 0;  
            a_col_loop: for (int k = 0; k < A_COLS; k++) {  
                sum_mult += in_a[i][k] * in_b[k][j];  
            }  
            out_c[i][j] = sum_mult;  
        }  
    }  
}
```

DSP48s: 48
Latency: 263 clocks
Loop II: 1

Pipelined Floating Point Matrix Multiply

```
void mm_pipelined_float(float in_a[A_ROWS][A_COLS],  
                         float in_b[A_COLS][B_COLS],  
                         float out_c[A_ROWS][B_COLS])  
{  
    float sum_mult;  
  
    // matrix multiplication of a A*B matrix  
    a_row_loop: for (int i = 0; i < A_ROWS; i++) {  
        b_col_loop: for (int j = 0; j < B_COLS; j++) {  
            sum_mult = 0.0;  
            a_col_loop: for (int k = 0; k < A_COLS; k++) {  
                #pragma HLS pipeline  
                sum_mult += in_a[i][k] * in_b[k][j];  
            }  
            out_c[i][j] = sum_mult;  
        }  
    }  
}
```

DSP48s: 1
Latency: 18453 clocks
Loop II: 4

Pipelined Floating Point Matrix Multiply

```
void mm_pipelined_float_interchanged(float in_a[A_ROWS] [A_COLS] ,
                                      float in_b[A_COLS] [B_COLS] ,
                                      float out_c[A_ROWS] [B_COLS])
{
    float sum_mult[B_COLS];

    // matrix multiplication of a A*B matrix
    a_row_loop: for (int i = 0; i < A_ROWS; i++) {
        a_col_loop: for (int k = 0; k < A_COLS; k++) {
            b_col_loop: for (int j = 0; j < B_COLS; j++) {
                #pragma HLS pipeline
                float last = (k==0) ? 0.0 : sum_mult[j];
                float result = last + in_a[i][k] * in_b[k][j];
                sum_mult[j] = result;
                if(k == (A_COLS-1)) out_c[i][j] = result;
            }
        }
    }
}
```

DSP48s: 1

BRAM: 1

Latency: 4105 clocks

Loop II: 1

18-bit Parallel Dot-Product Matrix Multiply

```
#include <ap_int.h>
void mm_18_parallel_dot_product(ap_int<18> in_a[A_ROWS][A_COLS],
                                ap_int<18> in_b[A_COLS][B_COLS],
                                ap_int<18> out_c[A_ROWS][B_COLS])
{
    #pragma HLS ARRAY_PARTITION DIM=2 VARIABLE=in_a complete
    #pragma HLS ARRAY_PARTITION DIM=1 VARIABLE=in_b complete
    ap_int<18> sum_mult;

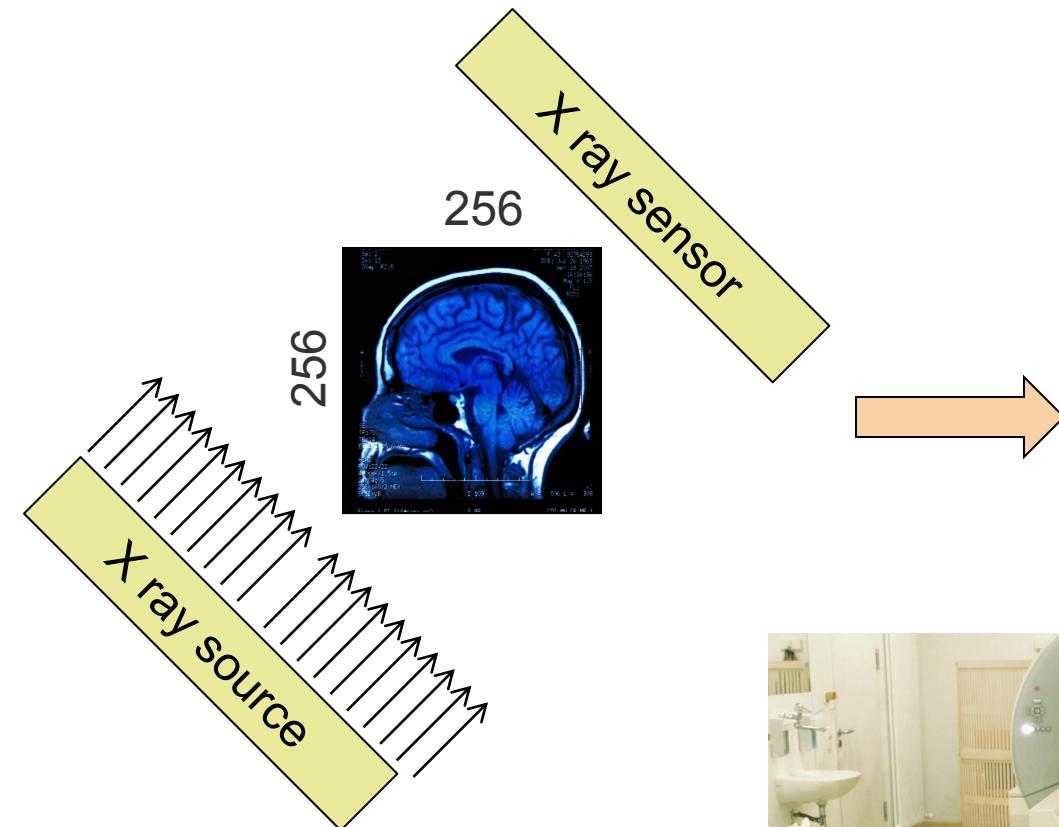
    // matrix multiplication of a A*B matrix
    a_row_loop: for (int i = 0; i < A_ROWS; i++) {
        b_col_loop: for (int j = 0; j < B_COLS; j++) {
            #pragma HLS pipeline
            sum_mult = 0;
            a_col_loop: for (int k = 0; k < A_COLS; k++) {
                sum_mult += in_a[i][k] * in_b[k][j];
            }
            out_c[i][j] = sum_mult;
        }
    }
}
```

DSP48s: 16
Latency: 260 clocks
Loop II: 1



Zynq Accelerated Applications

X-Ray Tomography Scanning



$$\text{sinoSize} = 256 * \sqrt{2}$$

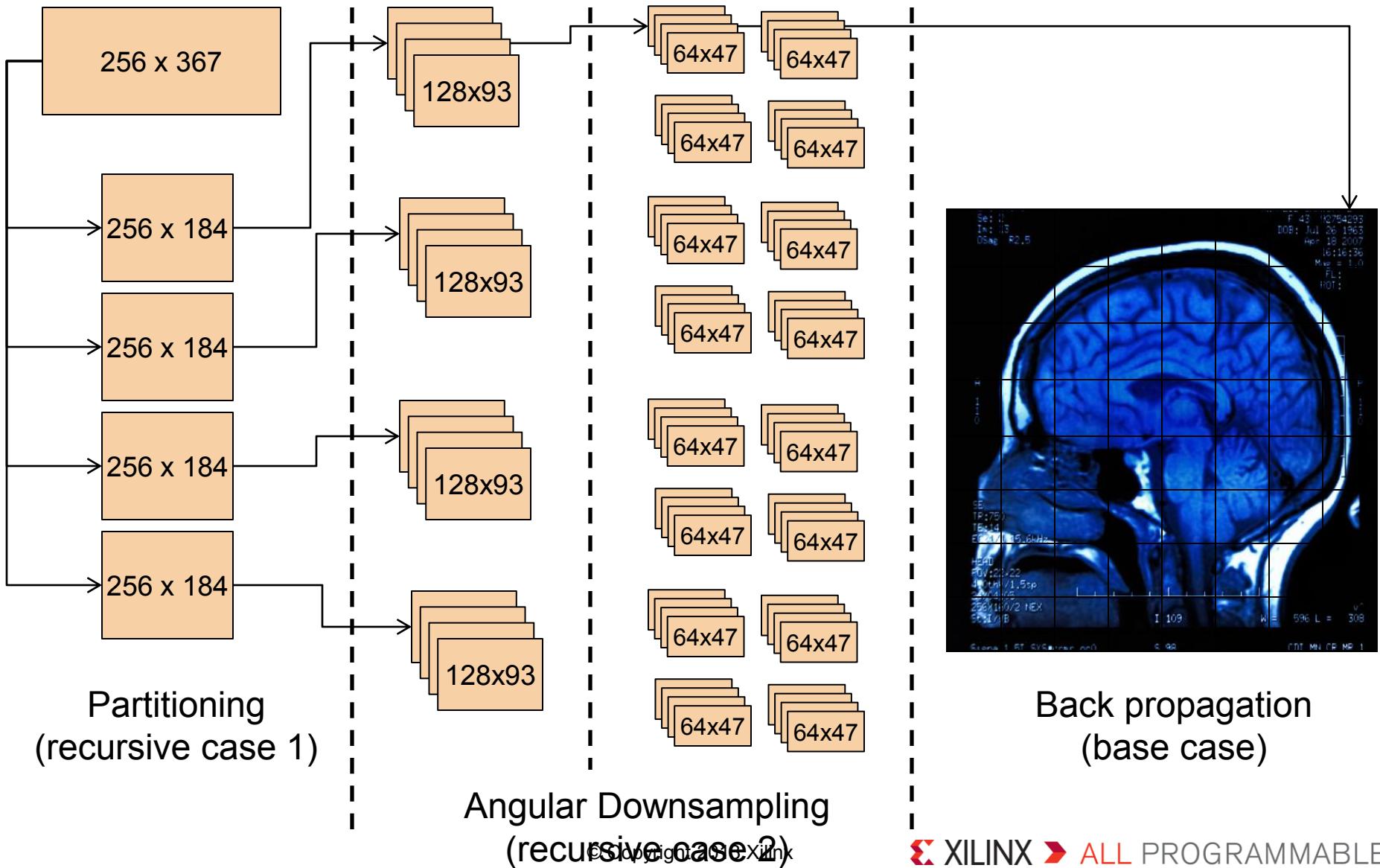
$$\begin{aligned}\text{sinoNum} &= \\ \text{angles} &= \\ 256\end{aligned}$$

256 x 367



[Video](#)

Backprojection Algorithm structure



Application characteristics

- **Total dataset will not easily fit in blockram or cache**

- $256 \times 367 \times 32\text{bit} = 275 \text{ KByte}$
- Recursive case 1 is 2x input size
- Recursive case 2 is same as input size

- **Downsampling reduces overall operations**

- Each downsampling stage reduces operations by factor 2.

- **Partitioning and Downsampling improves memory locality**

- Output data sets are smaller

- **Partitioning and Downsampling partitions data sets**

- Output data sets can be processed in parallel

Backprojection Application

- **Open Source Linux-based Application**

- Compiles directly on ARM/Zynq
- Single-precision floating point

- **Lots of things that are not synthesizable**

- Memory allocation
- File I/O
- Recursion

Code Structure

```
bp(sino, size, tau, img)
    if(size < limit) {
        direct(sino, size, tau, img);           // Base case
    } else foreach quadrant {
        newSino = allocSino(newSinoSize, newNumSino);
        if(condition) {                      // With downsampling
            newSinoForNextIter(newSino, sino, newSinoSize);
        } else {                            // No downsampling
            newSinoForNextIter2(newSino, sino, newSinoSize);
        }
        subImage = getTile(img, quadrant);
        bp(newSino, newSize, tau, subImage); // Recursion
    case
        freeSino(newSino);
    }
```

Code Structure: HLS limitations

- Pointers must point to statically allocated structures
- Pointers to pointers must be inlined

```
typedef struct{
    int size;
    myFloat **pixel; // [size] [size]
} image;

typedef struct{
    int num;           // number of angles
    int size;          // length of each filtered sinograms
    myFloat T;
    myFloat **sino;   // [size] [size];
    myFloat *sine;    // [size];
    myFloat *cosine;  // [size];
} sinograms;
```

Acceleration Approach

➤ Two functions with HLS-generated accelerators

- ap_newSinoForNextIter
 - Decomposes a sinogram into smaller sinogram tiles, with angular downsampling
- ap_direct
 - Computes result for a tile of the output image from a sinogram tile.

➤ Most code runs on ARM

- Memory allocation
- File I/O
- Sinogram Decomposition without angular downsampling

➤ Pipelined Coherent DMA

- Use good tile processing order for data locality

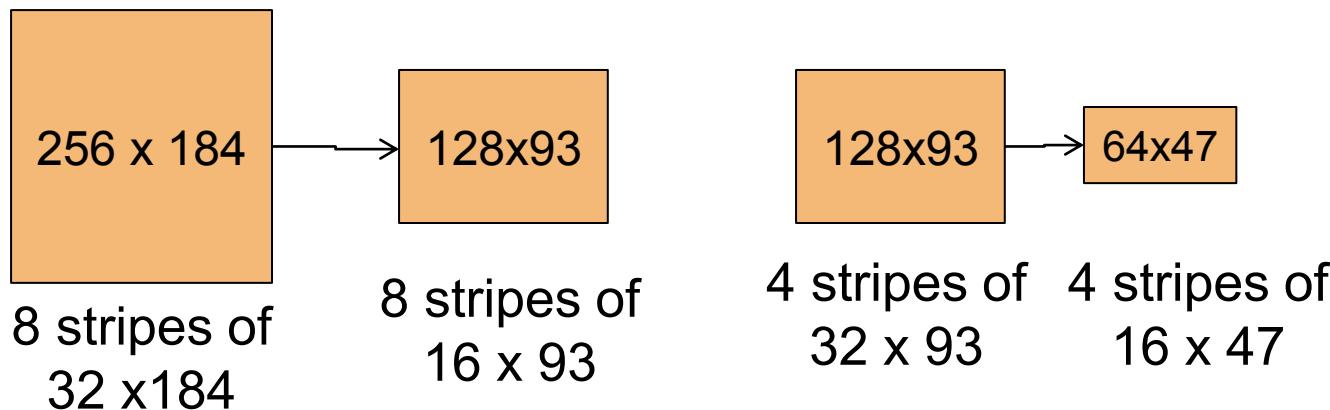
Code Transformations

► **ap_direct: 1 hour**

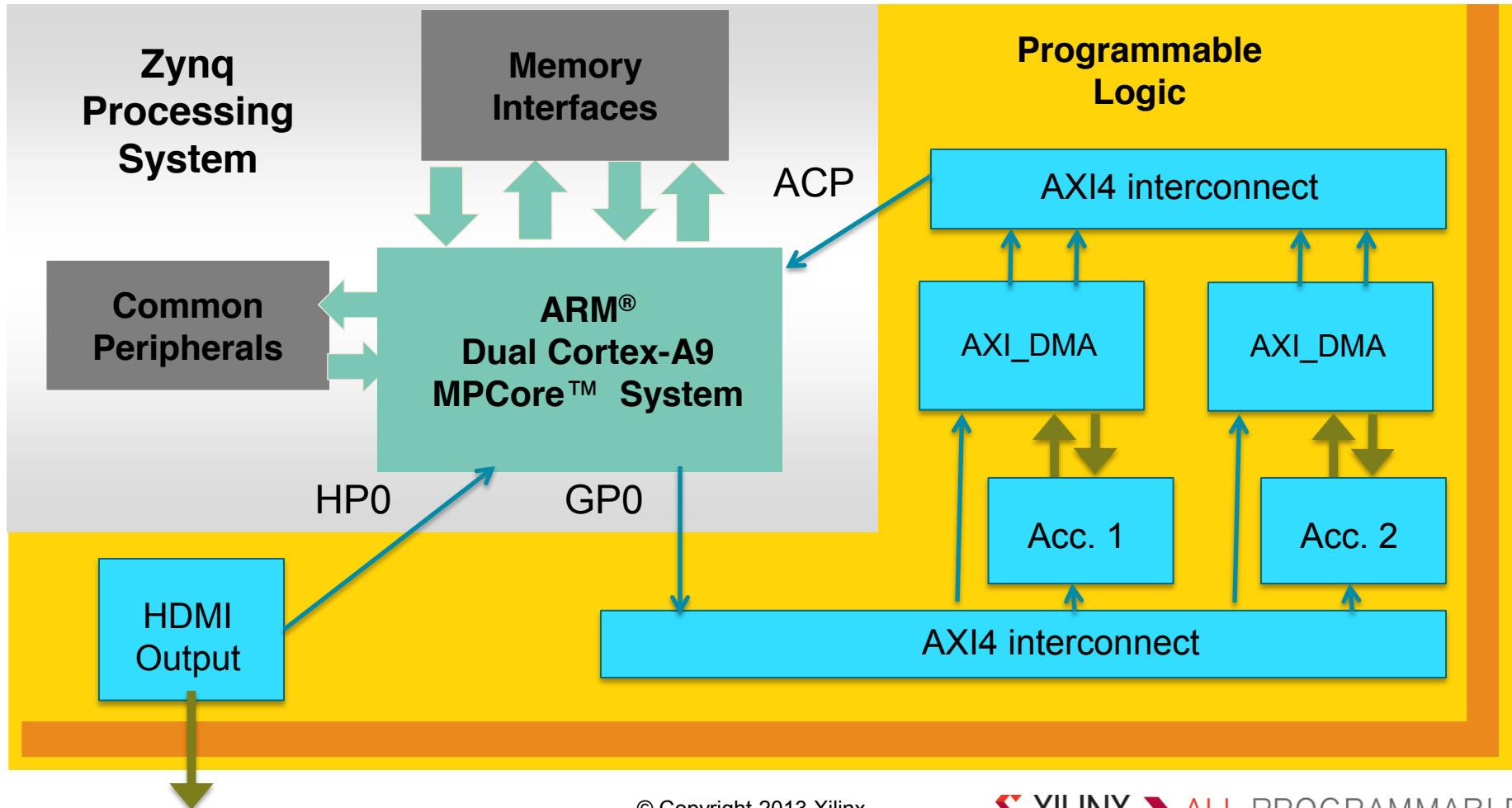
- Introduce statically allocated buffers to resolve pointers.

► **ap_newSinoForNextIter: 4 hours**

- Introduce statically allocated buffers to resolve pointers.
- Stripe-based processing of large data set (loop refactoring)



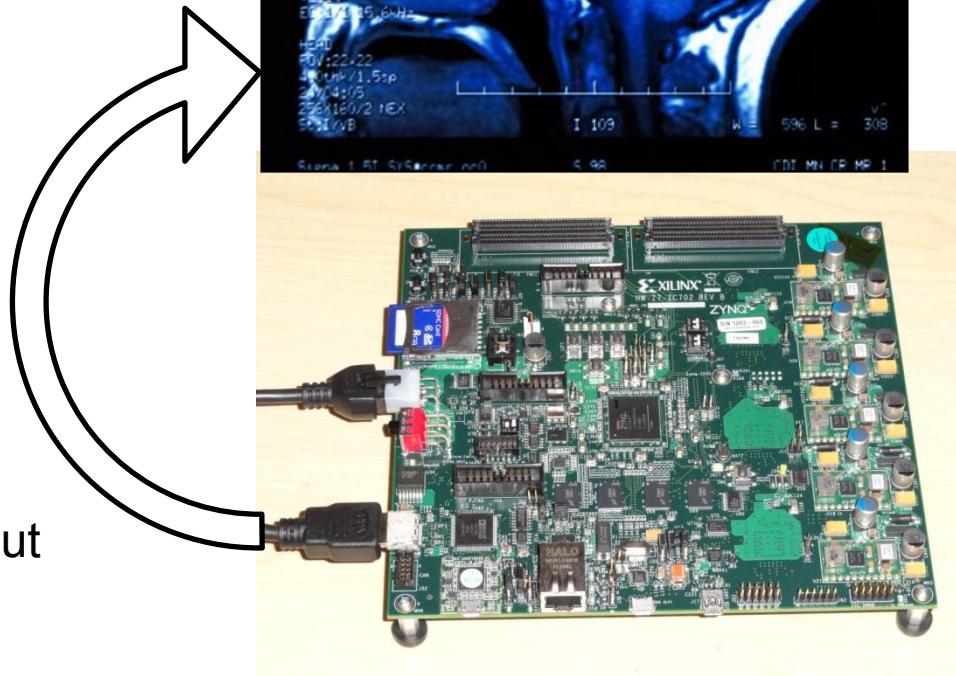
Architecture (With Coherent DMA)



Board + video

Left side of the screen:
SW running, ~ 1.5 fps

Right side of the screen:
SW+ HW running, ~ 10 fps



Intelligent Vision Applications for FPGAs



HD
Surveillance



Driver
Assistance



Video
Conferencing



Machine
Vision



A&D UAV

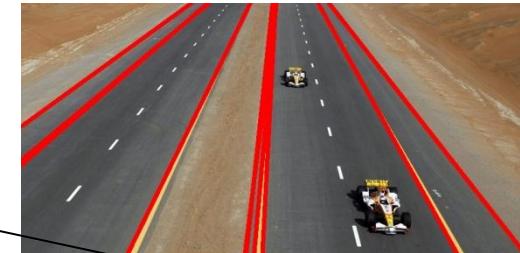


Office-class
MFP

Lane Detection – Algorithm Overview

► Lane Detection

- Analyze a video frame to detect road lane markings



RGB to
Gray Conversion



Image
Histogram
Equalization



Edge
Detection



Hough
Transform

Lane Detection

Application characteristics

- **Total dataset will not easily fit in blockram or cache**
 - $1920 \times 1080 \times 32$ bit = ~8 MB
- **Predictable access patterns and algorithms with high spatial locality**
 - line buffers and frame buffers
- **Applications are heterogeneous**
 - Pixel processing (good for FPGA)
 - Frame-level processing (good for processor)

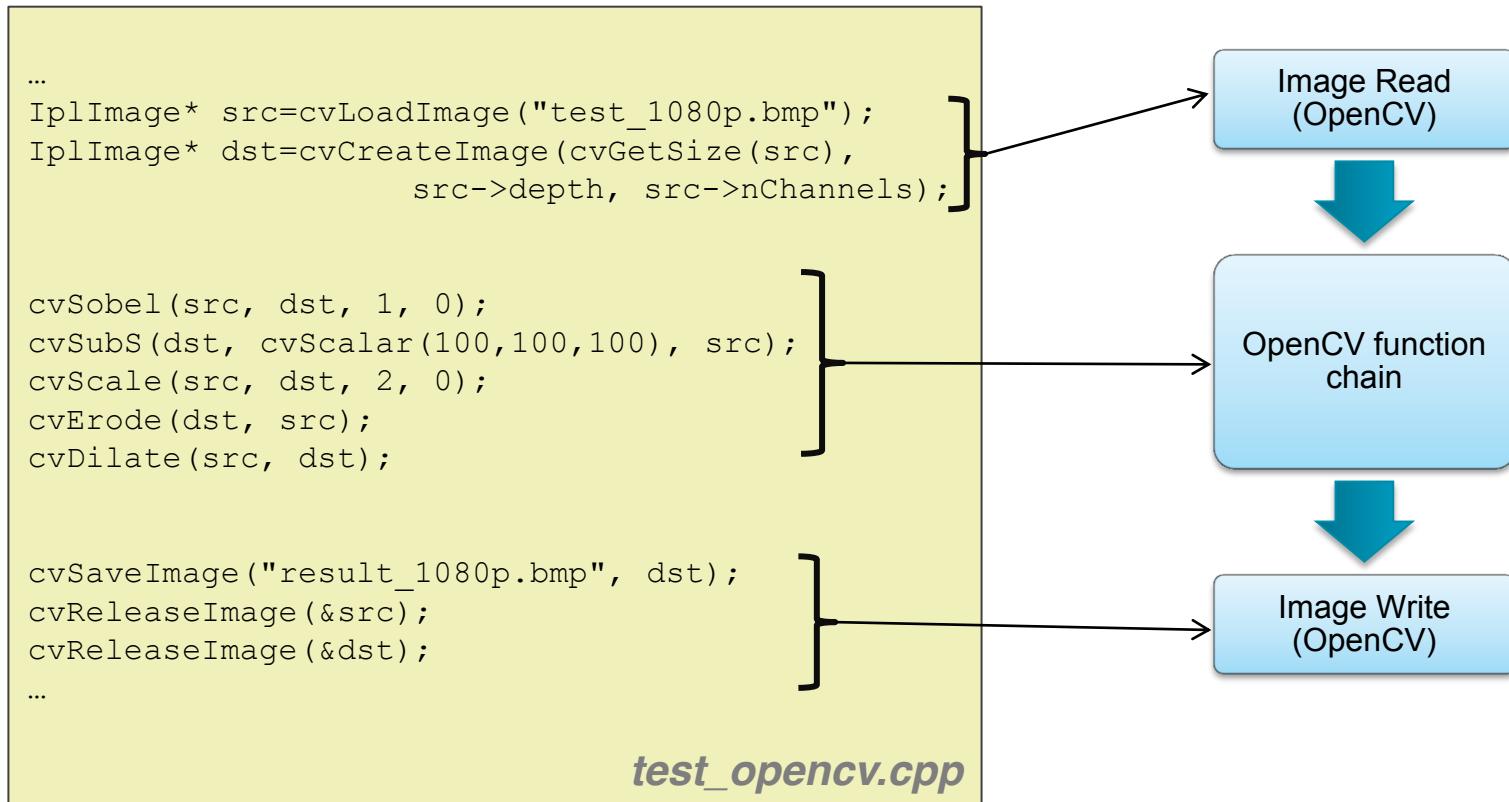
Acceleration Approach

- **Pixel processing accelerators with deep dataflow pipelines**
 - Video Function library corresponding to OpenCV functions
 - Extract features from pixels
- **Frame rate processing runs on ARM**
 - UI
 - Feature matching
 - Decision Making
- **Pipelined High performance DMA for video**
- **Features through general purpose interfaces**

OpenCV Code

➤ One image input, one image output

- Processed by chain of functions sequentially



Accelerated with Vivado HLS video library

➤ Top level function extracted for HW acceleration

```
#include "hls_video.h" // header file of HLS video library
#include "hls_opencv.h" // header file of OpenCV I/O

// typedef video library core structures
typedef hls::AXI_Base<32> AXI_PIXEL;
typedef hls::stream<AXI_PIXEL> AXI_STREAM;
typedef hls::Scalar<3, uchar> RGB_PIXEL;
typedef hls::Mat<1080,1920,HLS_8UC3> RGB_IMAGE;

void top(AXI_STREAM& src_axi, AXI_STREAM& dst_axi,
         int rows, int cols);

#include "top.h"

...
IplImage* src=cvLoadImage("test_1080p.bmp");
IplImage* dst=cvCreateImage(cvGetSize(src),
                           src->depth, src->nChannels);

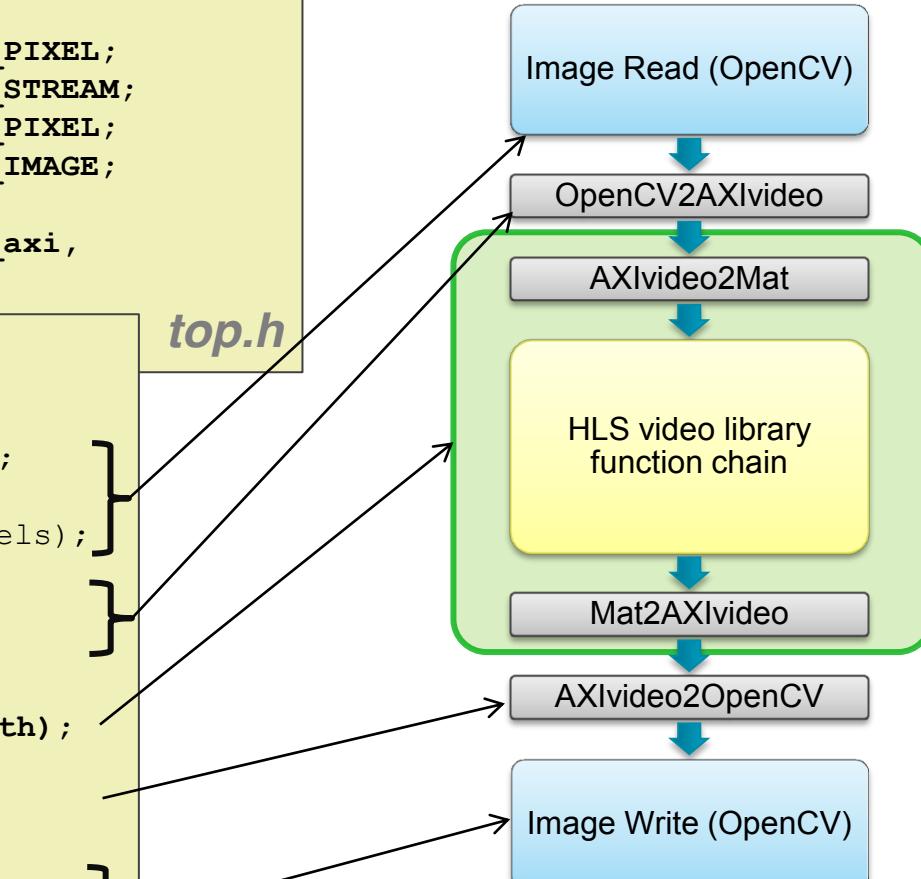
AXI_STREAM src_axi, dst_axi;
IplImage2AXIvideo(src, src_axi);

top(src_axi, dst_axi, src->height, src->width);

AXIvideo2IplImage(dst_axi, dst);

cvSaveImage("result_1080p.bmp", dst);
cvReleaseImage(&src);
cvReleaseImage(&dst);
```

test.cpp



Accelerated with Vivado HLS video library

➤ HW Synthesizable Block for FPGA acceleration

- Consists of video library function and interfaces
- Replace OpenCV function with similar function in hls namespace

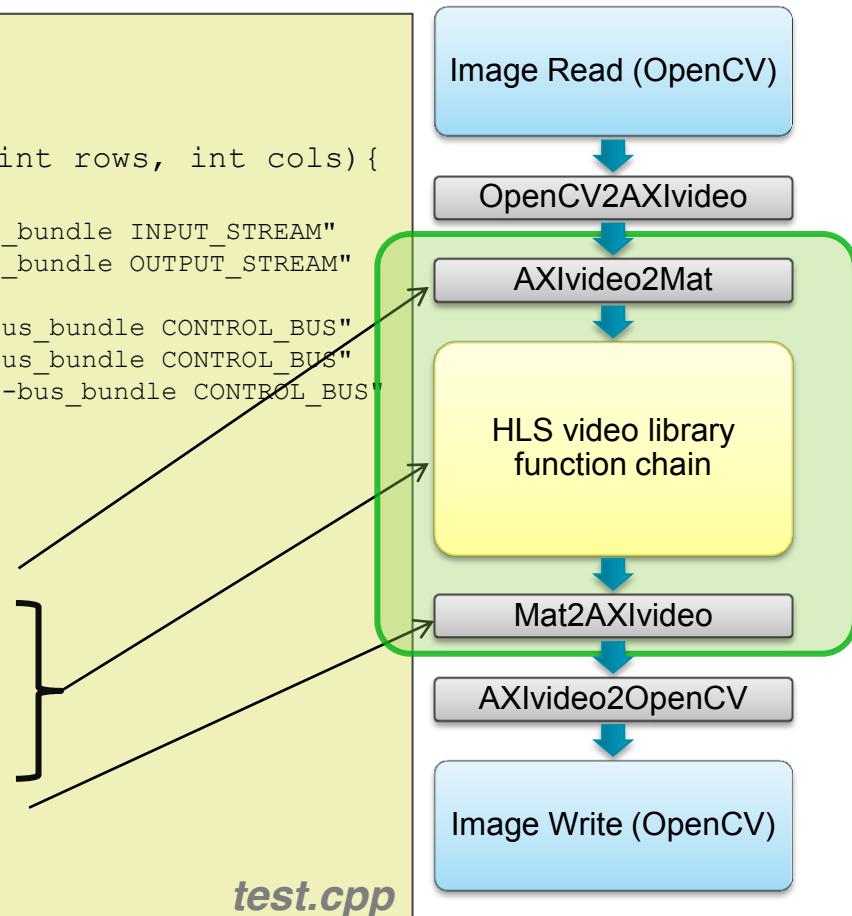
```
#include "top.h"
#include "ap_interfaces.h"

void top(AXI_STREAM& src_axi, AXI_STREAM& dst_axi, int rows, int cols){
    //Create AXI streaming interfaces for the core
#pragma HLS RESOURCE core=AXIS variable=src_axi metadata="-bus_bundle INPUT_STREAM"
#pragma HLS RESOURCE core=AXIS variable=dst_axi metadata="-bus_bundle OUTPUT_STREAM"

#pragma HLS RESOURCE core=AXI_SLAVE variable=rows metadata="-bus_bundle CONTROL_BUS"
#pragma HLS RESOURCE core=AXI_SLAVE variable=cols metadata="-bus_bundle CONTROL_BUS"
#pragma HLS RESOURCE core=AXI_SLAVE variable=return metadata="-bus_bundle CONTROL_BUS"

    RGB_IMAGE img[6];
    RGB_PIXEL pix(100,100,100);

#pragma HLS dataflow
    hls::AXIvideo2Mat(src_axi, img[0]);
    hls::Sobel(img[0], img[1], 1, 0);
    hls::SubS(img[1], pix, img[2]);
    hls::Scale(img[2], img[3], 2, 0);
    hls::Erode(img[3], img[4]);
    hls::Dilate(img[4], img[5]);
    hls::Mat2AXIvideo(img[5], dst_axi);
}
```



2012.4 Beta: Video Library Function List

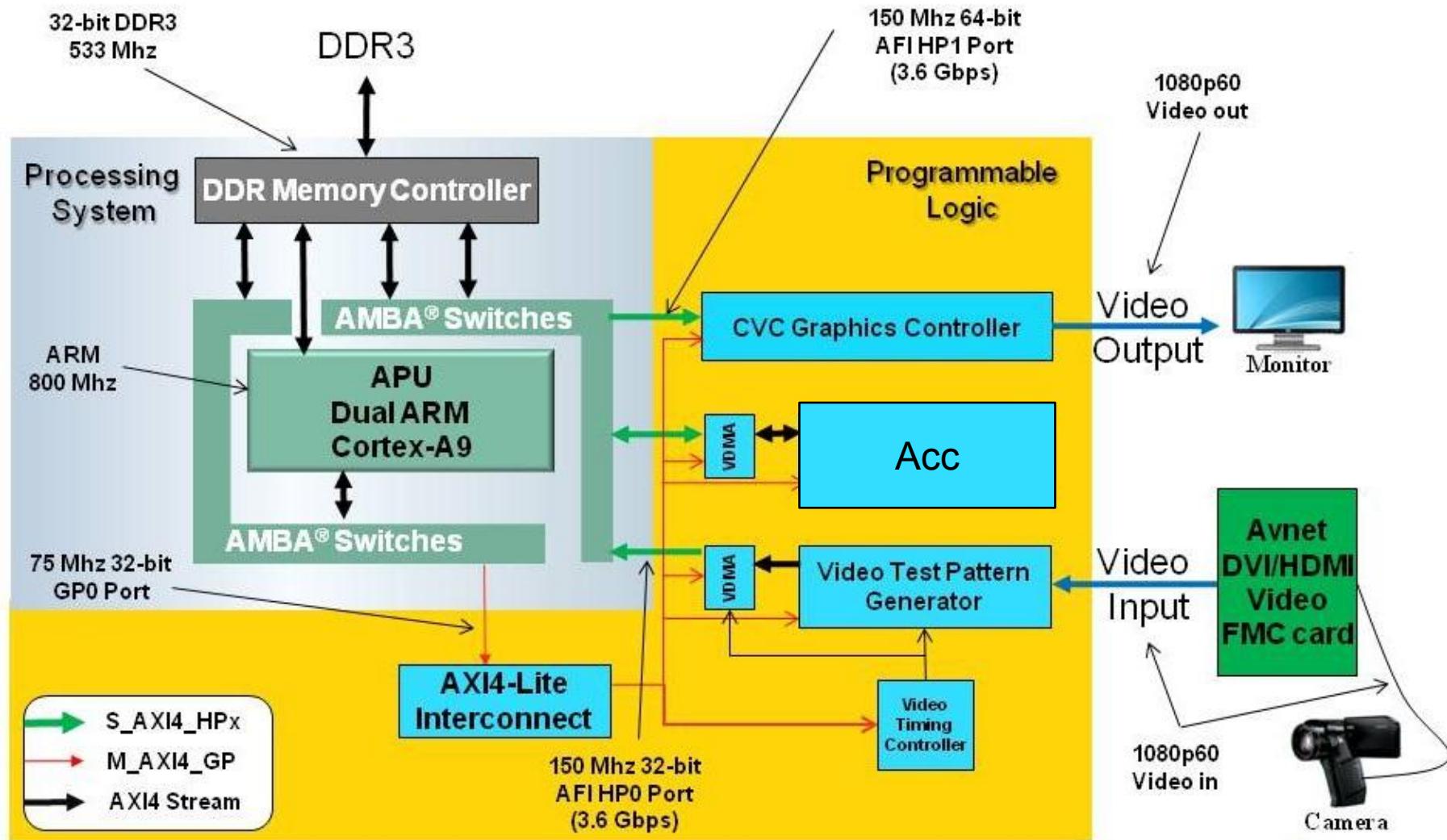
OpenCV I/O	cvMat2hlsMat IplImage2hlsMat CvMat2hlsMat
OpenCV I/O	hlsMat2cvMat hlsMat2IplImage hlsMat2CvMat
interfaces	hls::AXIvideo2Mat
interfaces	hls::Mat2AXIvideo
openCV basic function	hls::Filter2D
openCV basic function	hls::Erode
openCV basic function	hls::Dilate
openCV basic function	hls::Min
openCV basic function	hls::Max
openCV basic function	hls::MinS
openCV basic function	hls::MaxS
openCV basic function	hls::Mul
openCV basic function	hls::Zero
openCV basic function	hls::Avg

openCV basic function	hls::AbsDiff
openCV basic function	hls::CmpS
openCV basic function	hls::Cmp
openCV basic function	hls::And
openCV basic function	hls::Not
openCV basic function	hls::AddS
openCV basic function	hls::AddWeighted
openCV basic function	hls::Mean
openCV basic function	hls::SubRS
openCV basic function	hls::SubS
openCV basic function	hls::Sum
openCV basic function	hls::Reduce
openCV basic function	hls::Scale

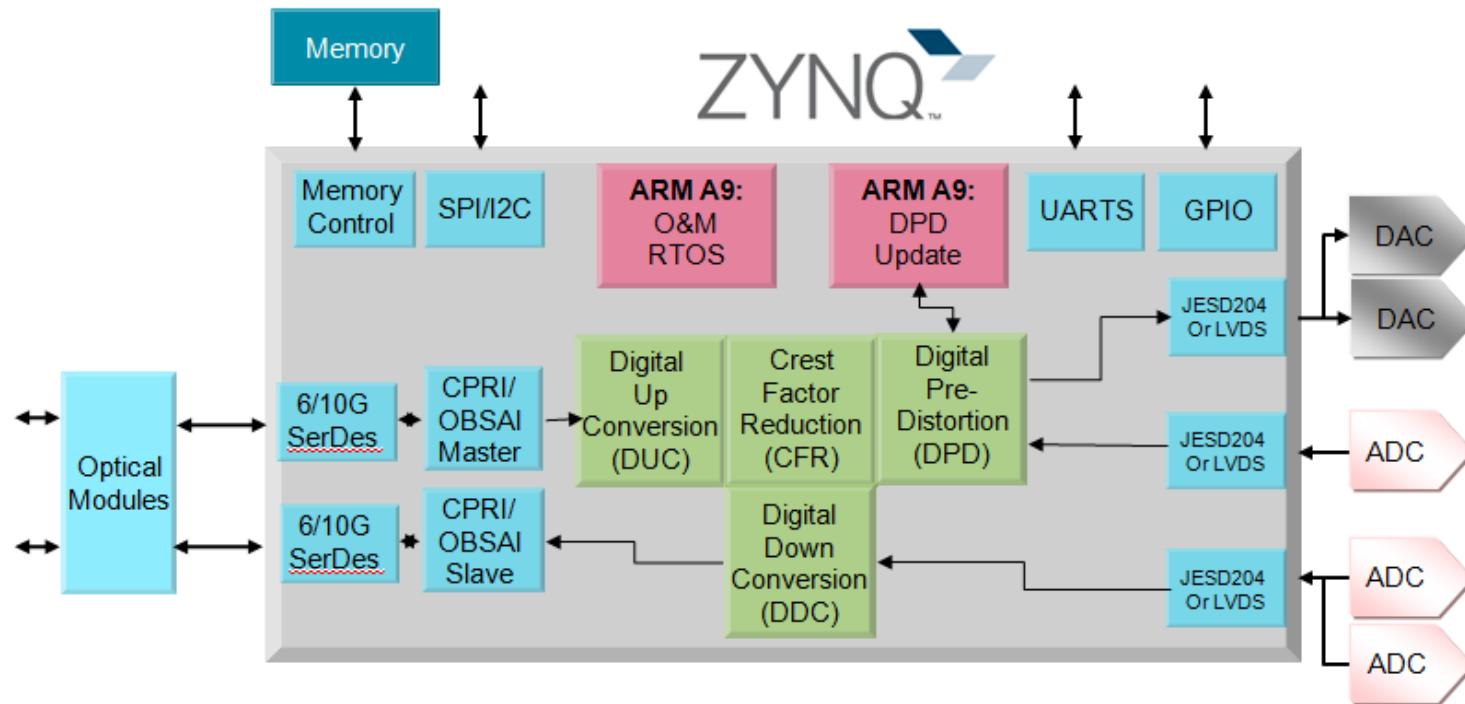
► For function signatures and descriptions, please refer to:

- Synthesizable functions in `hls_video.h`
- Interface functions in `hls_opencv.h`

Accelerator Architecture

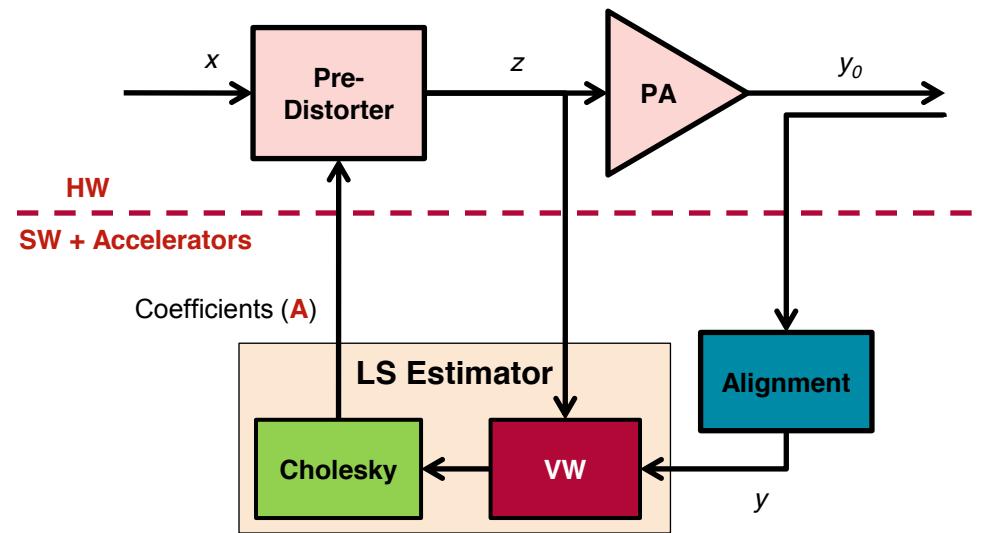


LTE Radio Digital Front End: Digital Pre-Distortion

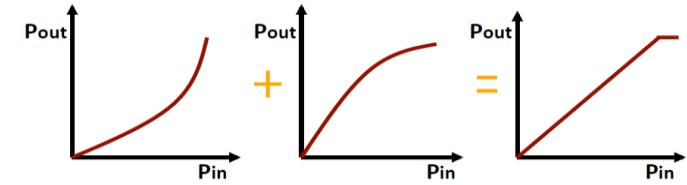


- Cost and power reduction by integrated solution
- Performance increase by exploiting the massive compute power of multi-core processors and programmable logic

Digital Pre-Distortion Functionality



- DPD negates PA non-linearity
 - PAs consume massive static power
 - DPD improves PA efficiency by ~35-40%



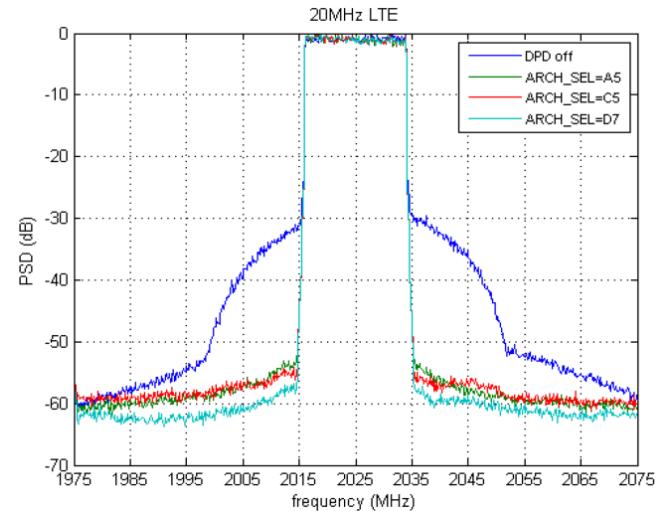
- Increase number of coefficients (K)
 - Better linearization, higher complexity

- Estimate pre-distorter coefficients (A):

$$Z = \begin{matrix} U(y) \\ \vdots \\ \vdots \end{matrix} \quad \begin{matrix} A \\ \vdots \\ \vdots \end{matrix}$$
$$(N \times 1) \quad (N \times K)$$

Diagram showing a vertical vector Z composed of rows U(y) and A, with dimensions (N x 1) and (N x K) respectively.

$$U^H Z = U^H U A$$
$$W = V A$$
$$V^{-1} W = A$$



Application characteristics

► Complex bare metal program

- Multiple loop nests, with no obvious bottleneck
- Fixed and floating point
- Complex numbers

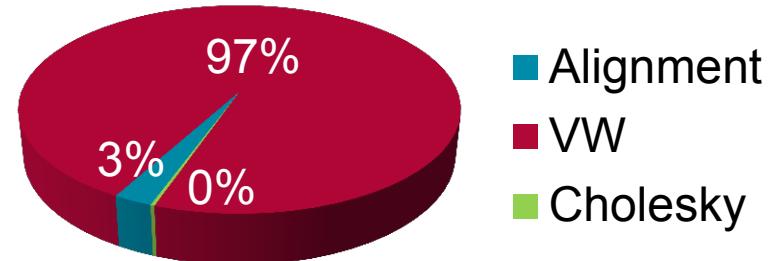
► Use software profiling

- Focus on VW functionality

► Initial Target: K = 64

- Look for speedup with minimal hardware usage

► More Speedup -> increase K



**Target Update Time: 300ms
(faster is better)**

	x86 2GHz	Zynq 800MHz
<i>Original</i>	0.66s	1.20s

Code Structure

```
for (int i = 0; i < NumCoeffs; ++i)
{
#pragma HLS pipeline II=2

    W[i].real +=
        (INT64) u[i].real*tx.real
        + (INT64) u[i].imag*tx.imag;

    W[i].imag +=
        (INT64) u[i].real*tx.imag
        - (INT64) u[i].imag*tx.real;
}
```

```
create_clock -period 5
set_part xc7z020clg484-2
```



Multipliers: 2
Adders: 2

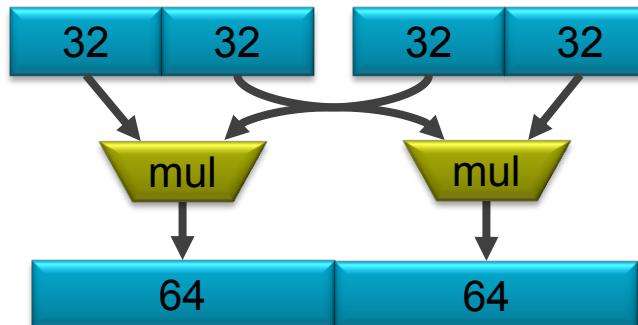
Software Optimization

➤ Use the right algorithm!

- Gains here are often easier than throwing hardware at the problem

➤ Use ARM NEON function intrinsics

- Low-level ARM-A9 programming



**Target Update Time: 300ms
(faster is better)**

	x86 2GHz	Zynq 800MHz
<i>Original</i>	0.66s	1.20s
<i>Optimized</i>	0.22s	0.54s
<i>With NEON</i>	n/a	0.25s



Speed-up: 5x

Using NEON Instructions

```
W[i].real += (INT64)uRow[i].real*sample.real  
+ (INT64)uRow[i].imag*sample.imag;
```



```
// load operands  
int32x2_t    tr = vdup_n_s32(sample.real);  
int32x2_t    ti = vdup_n_s32(sample.imag);  
int32x2x2_t  u  = vld2_s32((int32_t *) (uRow+i));  
int64x2_t    w  = vld1q_s64((int64_t *) (W+i));  
  
// do parallel computation  
w = vmlal_s32(w,u.val[0],tr);  
w = vmlal_s32(w,u.val[1],ti);  
  
// store result  
vst1q_s64((int64_t *) (W+i),w);
```

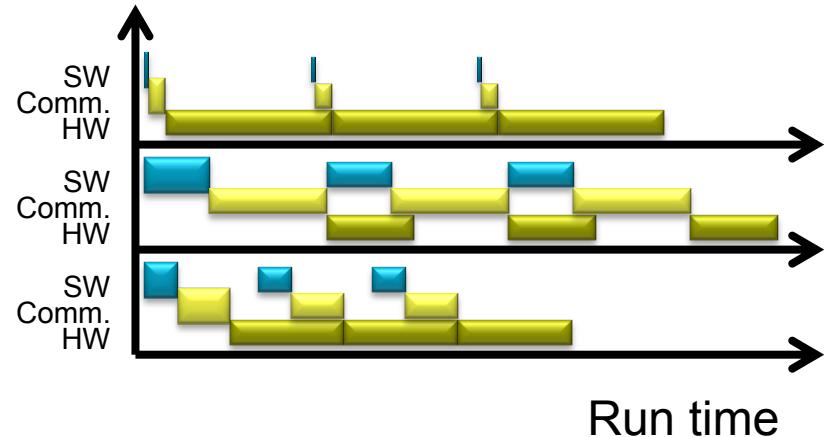
Acceleration Approach

- Neon intrinsics are OK, but HLS can do better

- with minimal code modification

- Pick right partitioning between processor code and accelerator

- Focus on efficient use of generated hardware
 - Tradeoff overall time and resources
 - time = sw + communication + hw
 - resources = communication + hw

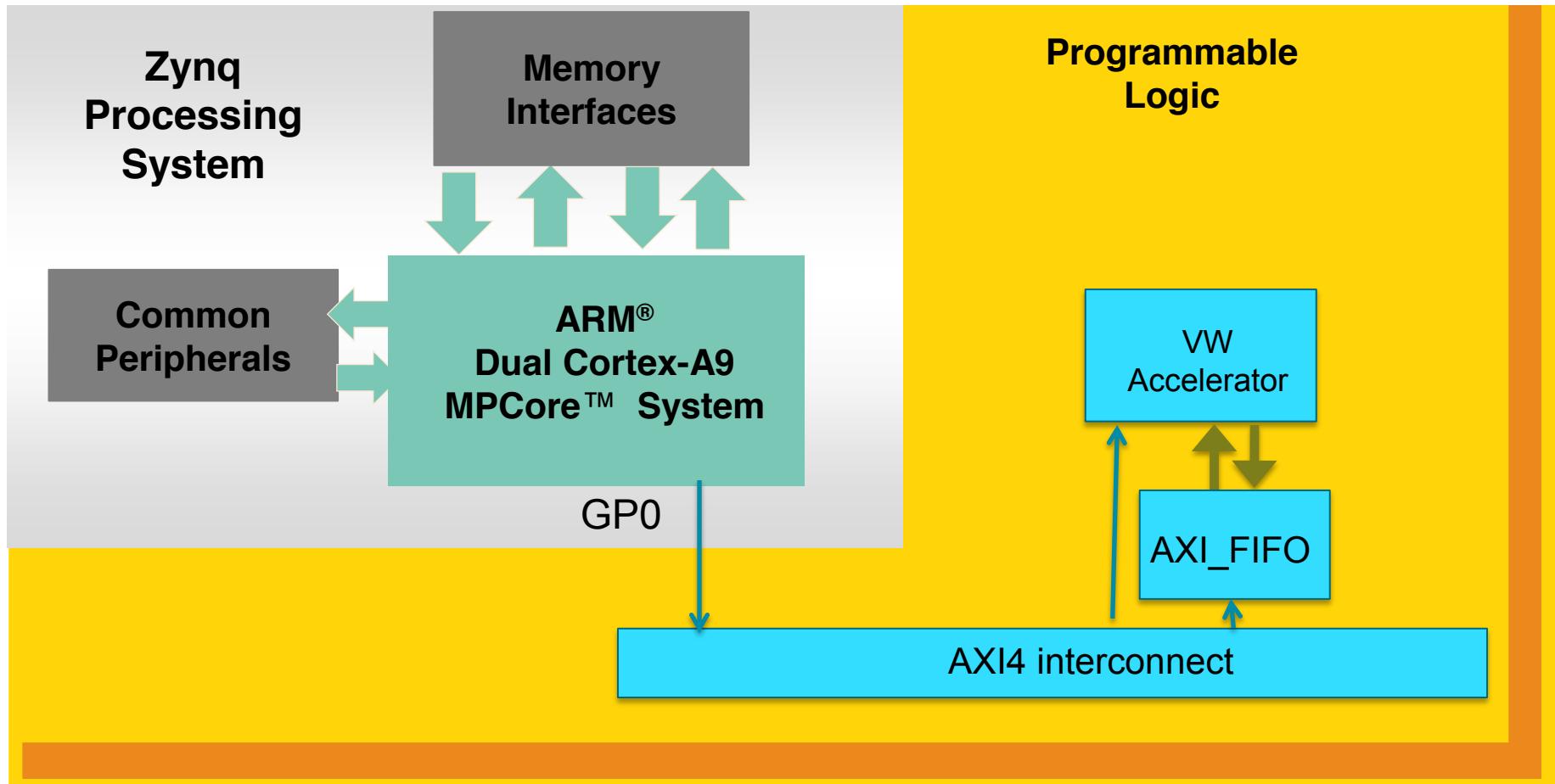


- Efficient memory-mapped IO with fifo

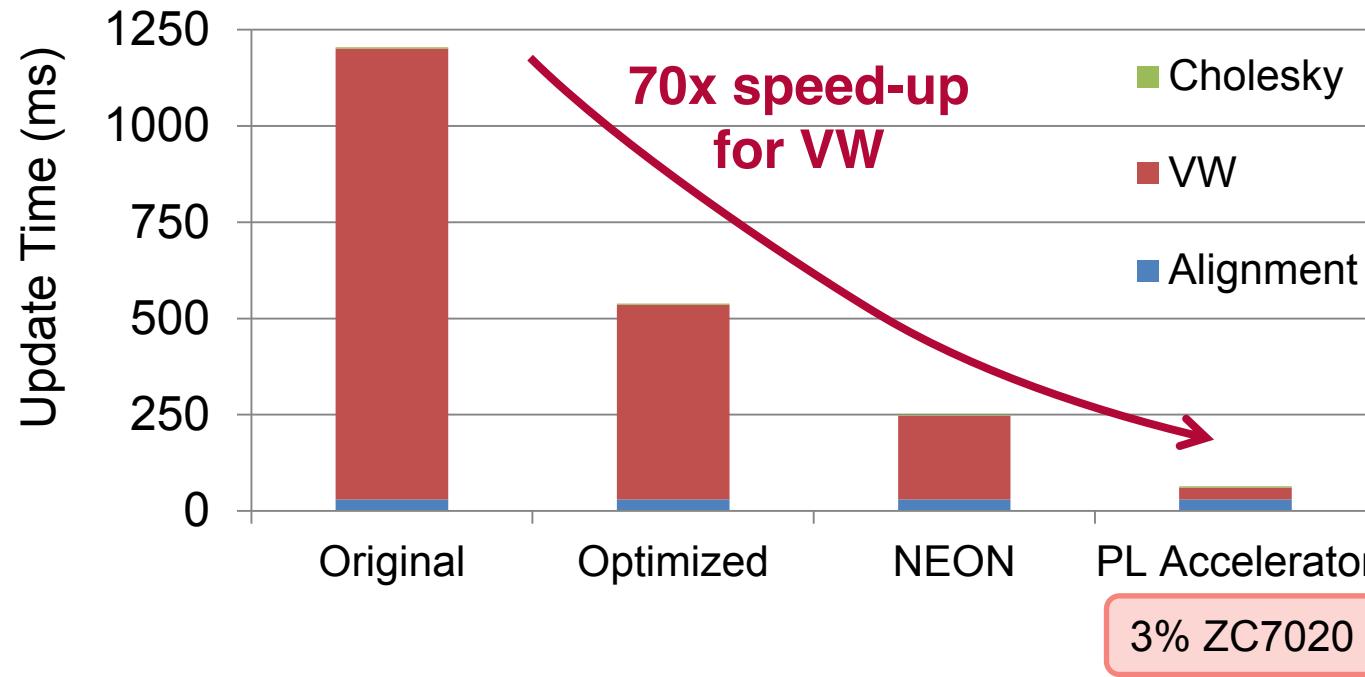
- DMA resources not justified

AXI FIFO Architecture

	FF	LUT
AXI Infrastructure	~300	~300
Accelerator	2552	2605



Digital Pre-Distortion on Zynq from C/C++



- Significant speed-up for existing designs
- OR: use speedup to solve bigger problems in same amount of time

Design exploration opportunities

- Maximize resource sharing

```
#pragma HLS allocation  
\\  
instances=mul limit=1 \\  
operation
```

- Insert pipelines

```
#pragma HLS pipeline II=1
```

- Vary number of coefficients

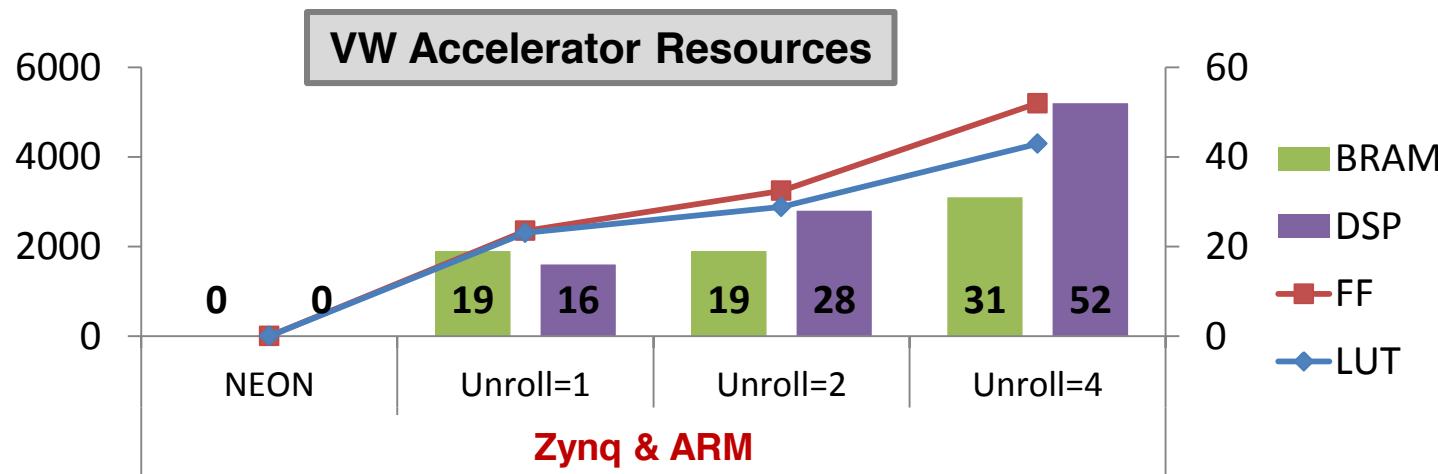
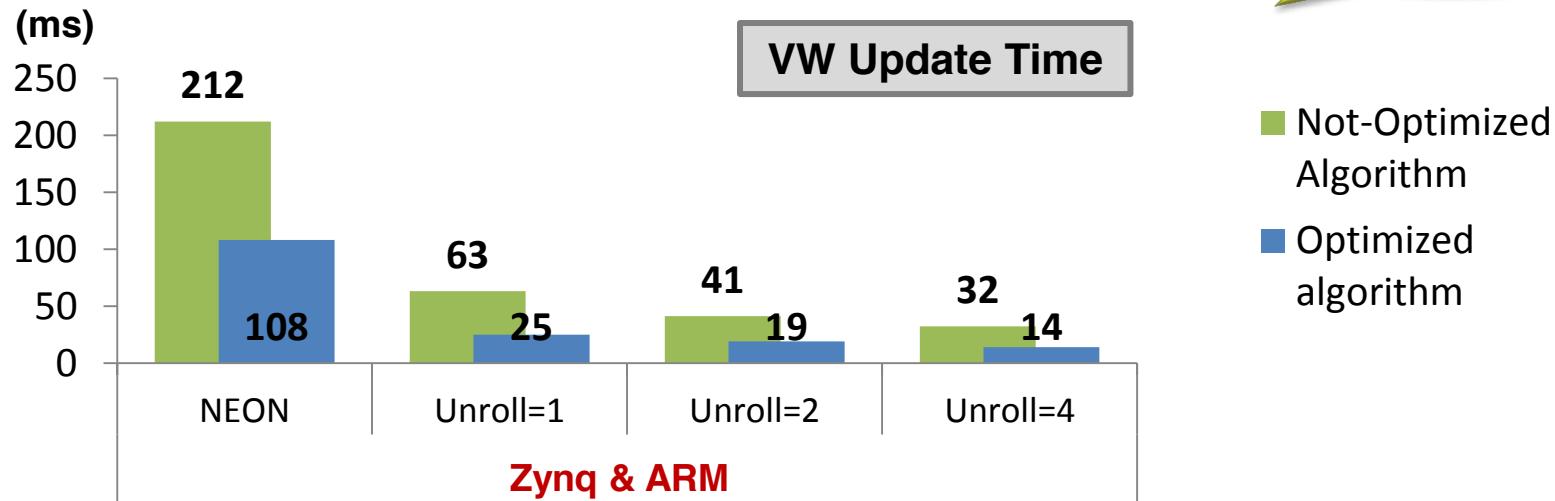
```
CINT64 W [MAX_COEFFS] ;
```

- Unroll loops to increase performance

```
#pragma HLS unroll \\  
factor=UNROLL_FACTOR
```

Accelerator Results

64 Coefficients



Conclusion

► Go forth and build!

► Many thanks to:

- Jan Langer
- Baris Ozgul
- Juanjo Noguera
- Kees Vissers
- Thomas Li
- Devin Wang
- Vinay Singh
- Duncan Mackay
- Dan Isaacs
- *And many others*