FPGA Field-Programmable Gate Arrays

Paolo Burgio paolo.burgio@unimore.it





Outline

- > Introduction to FPGAs
- > How to use them
- > Heterogeneous programming
- > FPGA-based heterogeneous programming
- > How to program it
- > Xilinx: now and soon...



The world, till now

(A)Symmetric multi-processing

Single or multi-core

Can be 1 bus, N busses, or any network

One or more cache levels

Main memory

CPU 2

TONE OF ONE OF MORE CACHE levels

One or more cache levels

I/O system



Reconfigurable Hardware

"Reconfigurable computing is intended to fill the gap between hardware and software, achieving potentially much higher performance than software, while maintaining a higher level of flexibility than hardware

(K. Compton and S. Hauck, Reconfigurable Computing: a Survey of Systems and Software, 2002)

Performance





Reconfigurable Hardware



Hardware

Flexibility



...eh?

We are used to have

- > On one (left) side, full programmable artifacts (software)
 - Run on single or multi-cores, designed for General Purpose computing
- > On one (right) side, hardware blocks to perform specific (subset of) operations

Performance









Hardware

Flexibility



What if?

...we had a "sea" of hardware blocks that we can program as we want

- > We can build cores
- > We can build co-processors
- > We can build what we want

What would you use them?

> For prototyping!

Hardware developent process is long and cumbersome

- > Imagine a full-fledged cores
- > Typically, years of development
- You can "try, and see whether it works"



History of reconfigurable echnologies

Logic gates (1950s-60s)

Regular structures for two-level logic (1960s-70s)

Muxes and decoders, PLAs

Programmable sum-of-products arrays (1970s-80s)

PLDs, complex PLDs

Programmable gate arrays (1980s-90s)

 densities high enough to permit entirely new class of application, e.g., prototyping, emulation, acceleration trend toward higher levels of integration



Field-programmable gate arrays

- "A field-programmable gate array (FPGA) is an integrated circuit designed to be configured by a customer or a designer after manufacturing."
- > Traditionally used for prototyping
 - Takes minutes vs years for "real" hardware
- > Tech has evolved so they are actively used in production settings
 - Less powerful, yet more energy-efficient than a GPU
 - Way more flexible

Integrated into System-on-chips

- > As reconfigurable accelerator
- > We'll see later...

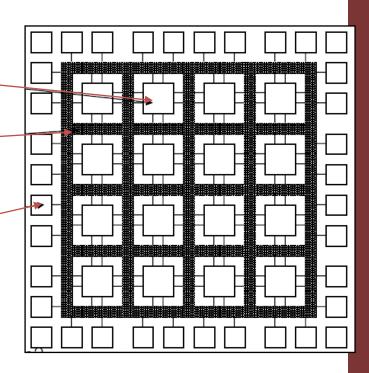


FPGAs

- > Logic blocks
 - to implement combinational and sequential logic
- > Interconnect
 - wires to connect inputs and outputs to logic blocks
- > I/O blocks
 - special logic blocks at periphery of device for external connections

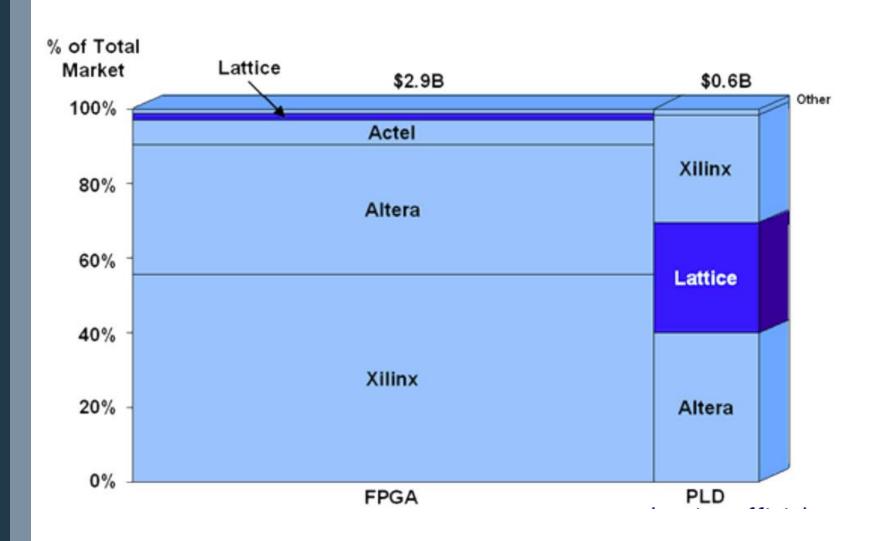
Key questions:

- > how to make logic blocks programmable?
- > how to connect the wires?
- > after the chip has been fabbed





Commercial FPGA companies

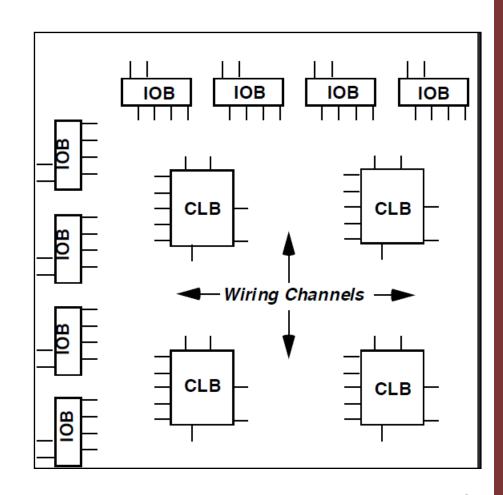




(Xilinx) Programmable Gate Arrays

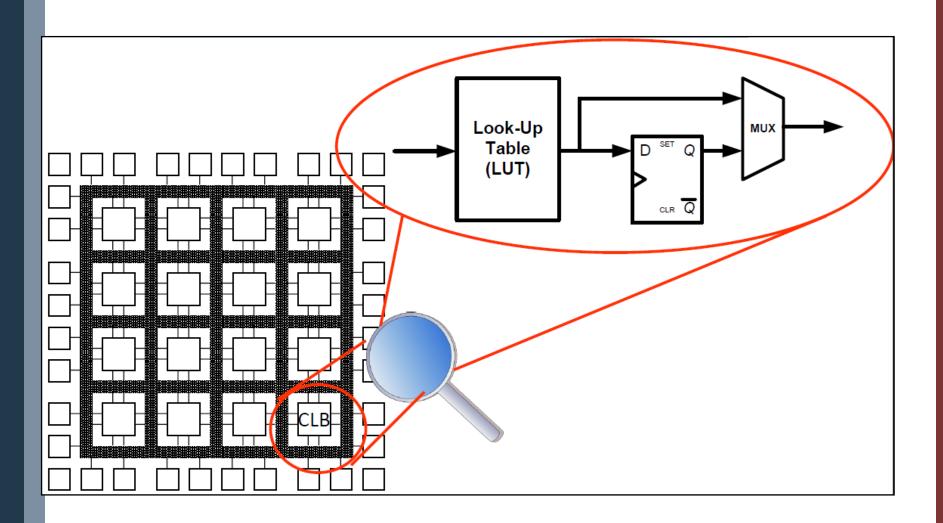
CLB - Configurable Logic Block

- > Built-in fast carry logic
- > Can be used as memory
- > Three types of routing
 - direct
 - general-purpose
 - long lines of various lengths
- > RAM-programmable
 - can be reconfigured





Simplified CLB Structure

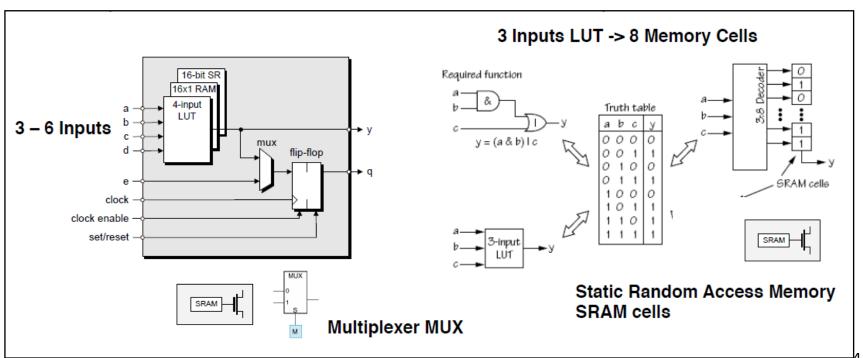




LookUp Tables

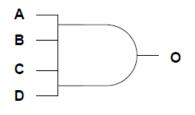
LUT contains Memory Cells to implement small logic functions

- > Each cell holds '0' or '1'
- > Programmed with outputs of Truth Table
- > Inputs select content of one of the cells as output

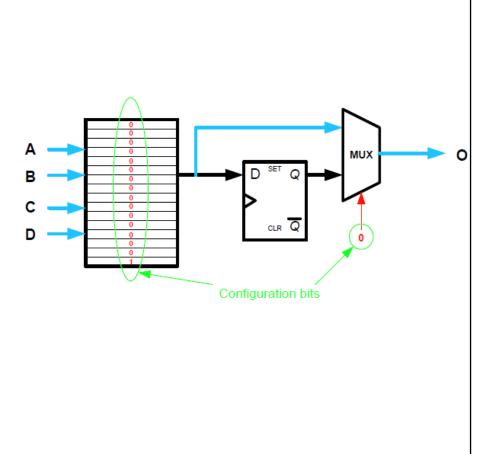




Example: 4-input AND gate

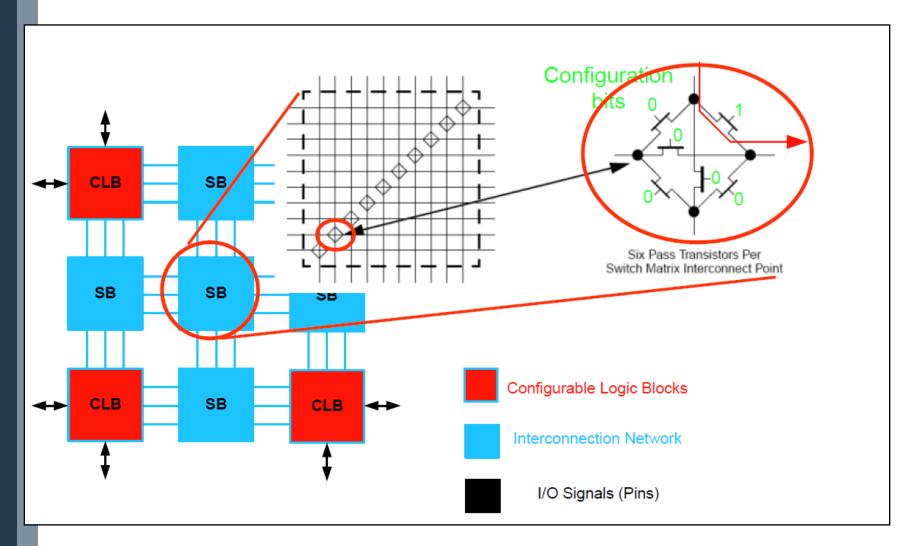


Α	В	С	D	0
0	0	0	0	0
0	0	0	1	0
0	0	1	0	0
0	0	1	1	0
0	1	0	0	0
0	1	0	1	0
0	1	1	0	0
0	1	1	1	0
1	0	0	0	0
1	0	0	1	0
1	0	1	0	0
1	0	1	1	0
1	1	0	0	0
1	1	0	1	0
1	1	1	0	0
1	1	1	1	1





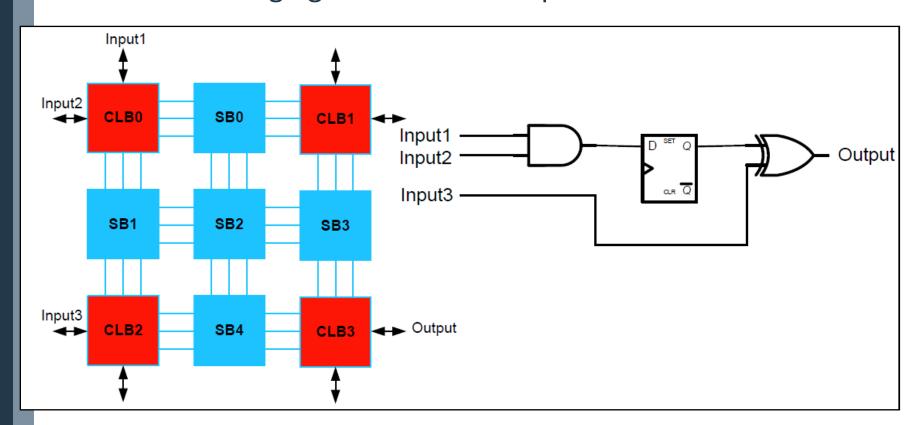
Interconnection Network





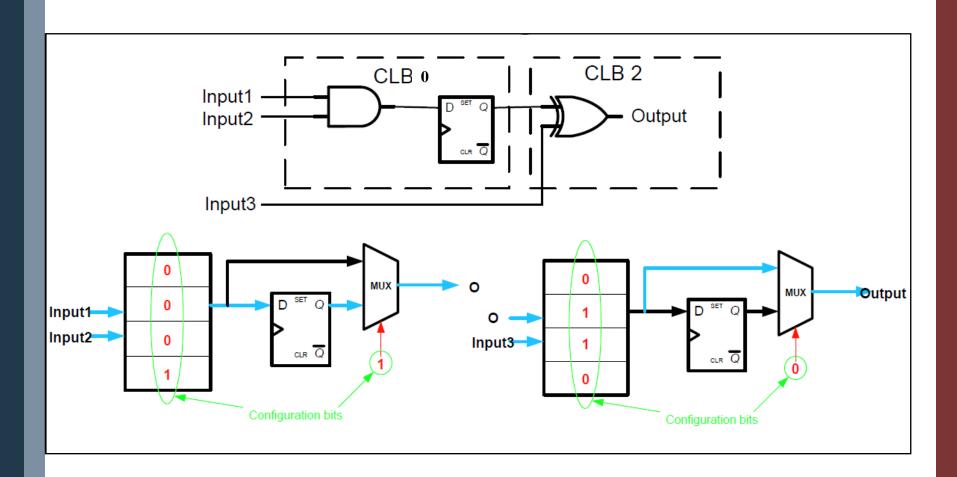
Example

> Determine the configuration bits for the following circuit implementation in a 2x2 FPGA, with I/O constraints as shown in the following figure. Assume 2-input LUTs in each CLB



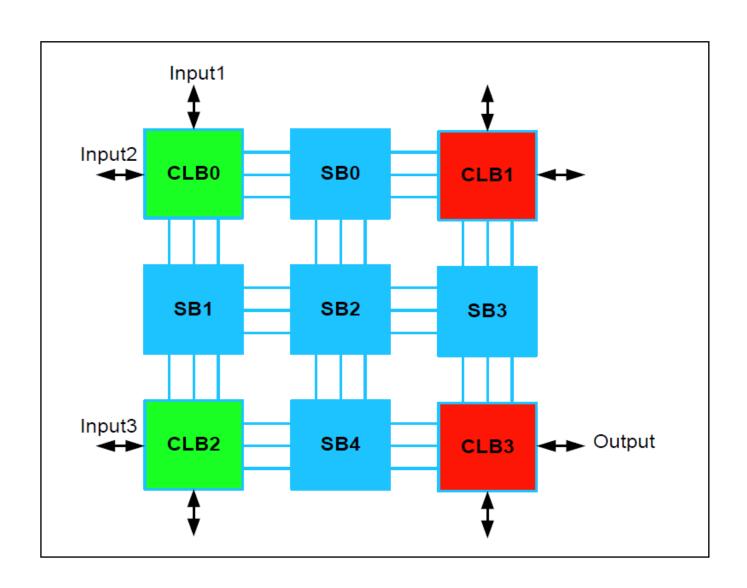


Configure CLBs



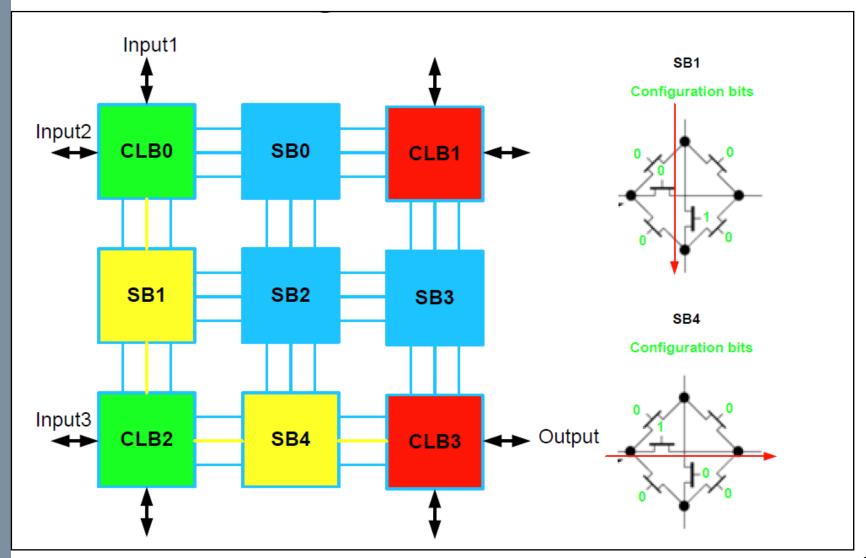


Placement: select CLBs





Routing: Select path





Configuration Bitstream

> The configuration bitstream must include ALL CLBs and SBs, even unused ones

> CLB0: 00011

> CLB1: ?????

> CLB2: 01100

> CLB3: XXXXX

> SB0: 000000

> SB1: 000010

> SB2: 000000

> SB3: 000000

> SB4: 000001



Some Definitions

- Object Code (aka "Bitstream): the executable active physical (either HW or SW) implementation of a given functionality
- Core: a specific representation of a functionality. It is possible, for example, to have a core described in VHDL, in C or in an intermediate representation (e.g. a DFG)
- > IP-Core: a core described using a HD Language combined with its communication infrastructure (i.e. the bus interface)
- Reconfigurable Functional Unit: an IP-Core that can be plugged and/or unplugged at runtime in an already working architecture
- Reconfigurable Region: a portion of the device area used to implement a reconfigurable core



Computer-Aided Design

Can't design FPGAs by hand!

- > way too much logic to manage, hard to make changes
- > Hardware description languages (HDL), es: Verilog, VHDL
 - specify functionality of logic at a high level
- > Logic synthesis
 - process of compiling HDL program into logic gates and flip-flops
- > Validation high-level simulation to catch specification errors
 - verify pin-outs and connections to other system components
 - low-level to verify and check performance



CAD Tool Path (cont'd)

> Technology mapping

map the logic onto elements available in the implementation technology (LUTs for Xilinx FPGAs)

> Placement and routing

- assign logic blocks to functions
- make wiring connections

> Partitioning and constraining

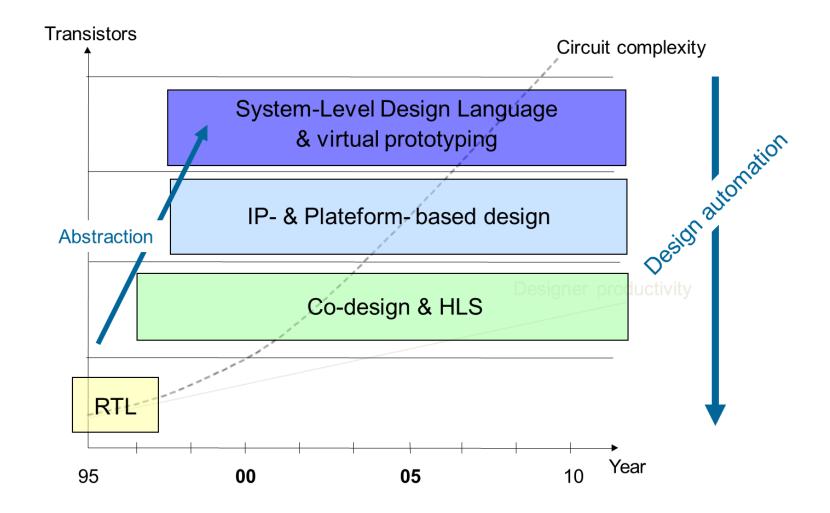
- if design does not fit or is unroutable as placed split into multiple chips
- if design it too slow prioritize critical paths, fix placement of cells, etc.
- few tools to help with these tasks exist today
- Generate programming files bits to be loaded into chip for configuration

High-level synthesis



Electronic System Level Design (ESLD)



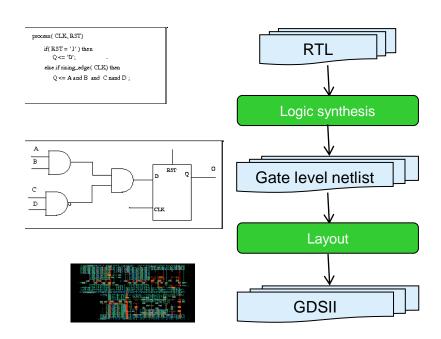




Typical HW design flow



 Starting from a Register Transfer Level description, generate an IC layout

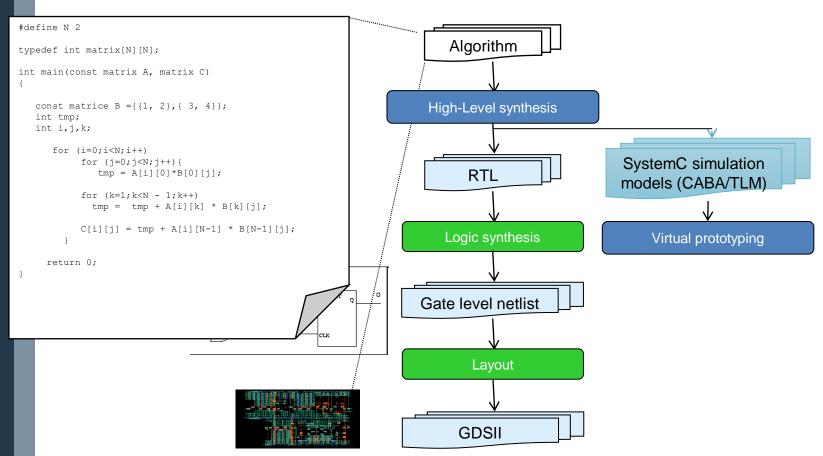




Typical HW design flow



 Starting from a Register Transfer Level description, generate an IC layout





High-Level Synthesis



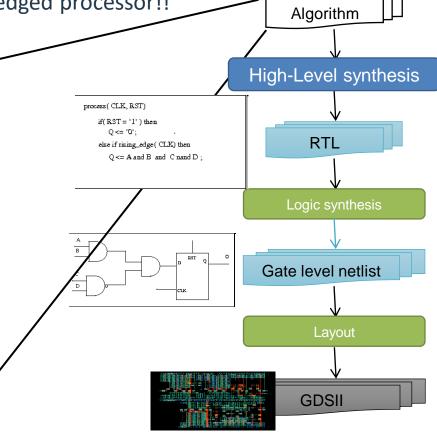
- > Starting from C code, generate RTL, and then, layout
 - Typically, HW accelerators (FFT?) Ips

Can also synthesize a fully fledged processor!!

Soft-cores

```
#define N 2
typedef int matrix[N][N];

int main(const matrix A, matrix C)
{
   const matrice B = {{1, 2}, { 3, 4}};
   int tmp;
   int i,j,k;
   for (i=0;i<N;i++)
      for (j=0;j<N;j++) {
       tmp = A[i][0]*B[0][j];
      for (k=1;k<N - 1;k++)
        tmp = tmp + A[i][k] * B[k][j];
       C[i][j] = tmp + A[i][N-1] *
            B[N-1][j];
   }
   return 0;
}</pre>
```





It is basically, a compiler!



> From C code

- Generates the "physical" representation of Hardware modules
- Registry Transfer Level, RTL
- That will be deployed on the board

> Automatically



Synthesizable C subset



> No pointers

- Statically unresolved
- Arrays are allowed!
- > No standard function call
 - printf, scanf, fopen, malloc...
- > Function calls are allowed
 - Can be in-lined or not.
- > Nearly all datatypes are allowd
 - Specific datatypes are encouraged
 - Bit accurate integers, fixed point, signed, unsigned...



Example #1: a simple C code

```
#define N 16
int main(int data in, int *data out) {
  static const int Coeffs [N] = \{ 98, -39, -327, 439, 950, -2097, -1674, 9883, 
                                   9883, -1674, -2097, 950, 439, -327, -39, 98 };
 int Values[N];
 int temp;
  int sample,i,j;
  sample = data in;
  temp = sample * Coeffs[N-1];
  for (i = 1; i \le (N-1); i++) {
   temp += Values[i] * Coeffs[N-i-1];
  for (j=(N-1); j>=2; j-=1)
   Values[j] = Values[j-1];
 Values[1] = sample;
  *data out=temp;
  return 0;
```



Example #2: bit accurate C++ code

```
#include "ac fixed.h" // From Mentor Graphics
#define PORT SIZE ac fixed<16, 12, true, AC RND, AC SAT> // 16 bits, 12 bits after the \
  point, quantization = rounding, overflow = saturation
#define N 16
int main(PORT SIZE data in, PORT SIZE &data out) {
  static const PORT SIZE Coeffs [N]= { 1.1, 1.5, 1.0, 1.0, 1.7, 1.8, 1.2, 1.0,
                                       1.6, 1.0, 1.5, 1.1, 1.9, 1.3, 1.4, 1.7 };
  PORT SIZE Values[N];
  PORT SIZE temp;
  PORT SIZE sample;
  sample= data in;
  temp = sample * Coeffs[N-1];
  for (int i = 1; i <= (N-1); i++) {
    temp = Values [i] * Coeffs[N-i-1] + temp;
  for (int j = (N-1); j \ge 2; j-=1) {
    Values[j] = Values[j-1];
 Values[1] = sample;
  data out=temp;
 return 0;
```



High-level transformations



> Loops

- Loop pipelining,
- Loop unrolling
- Loop merging
- Loop tiling
- ...

> Arrays mapping

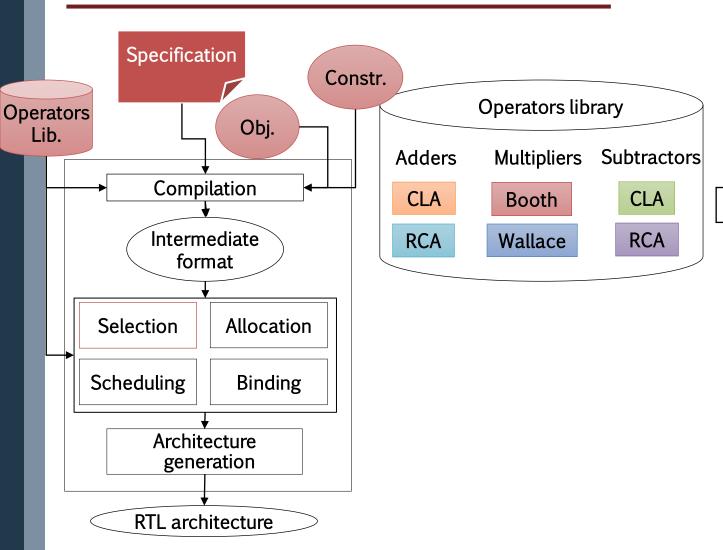
- Arrays can be mapped on memory banks
- Arrays can be synthesized as registers
- Constant arrays can be synthesized as logic
- **–** ...

> Functions

- Function calls can be in-lined
- Function is synthesized as an operator
 - > Sequential, pipelined, functional unit...
- Single function instantiation
- ..



HLS steps: Inputs

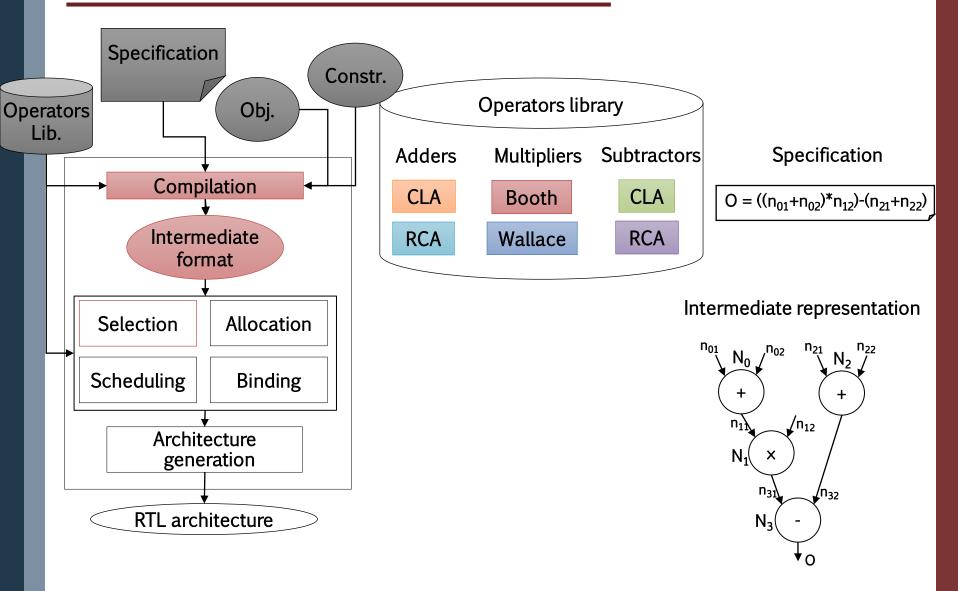


Specification

$$O = ((n_{01} + n_{02})^* n_{12}) - (n_{21} + n_{22})$$

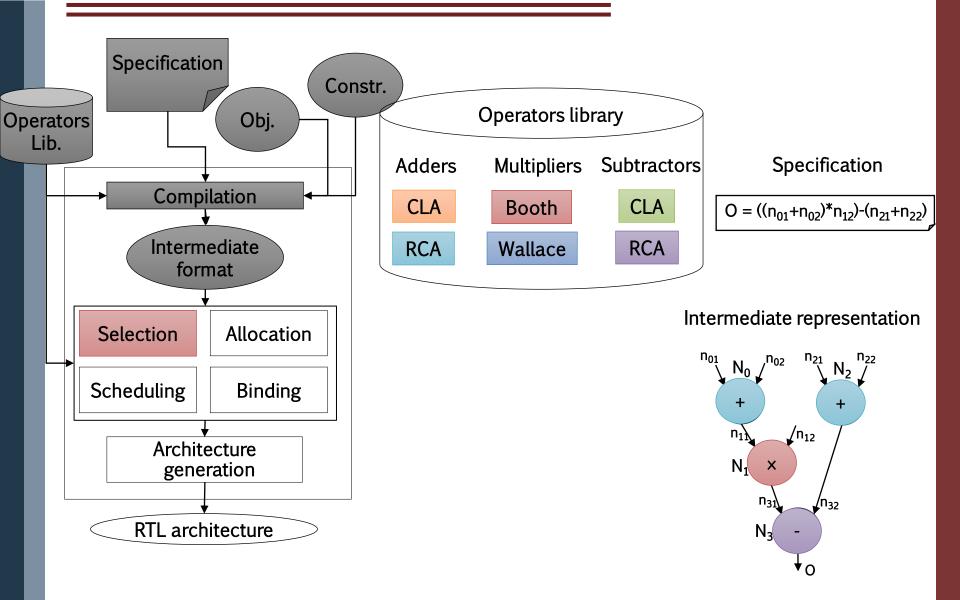


HLS steps: Compilation



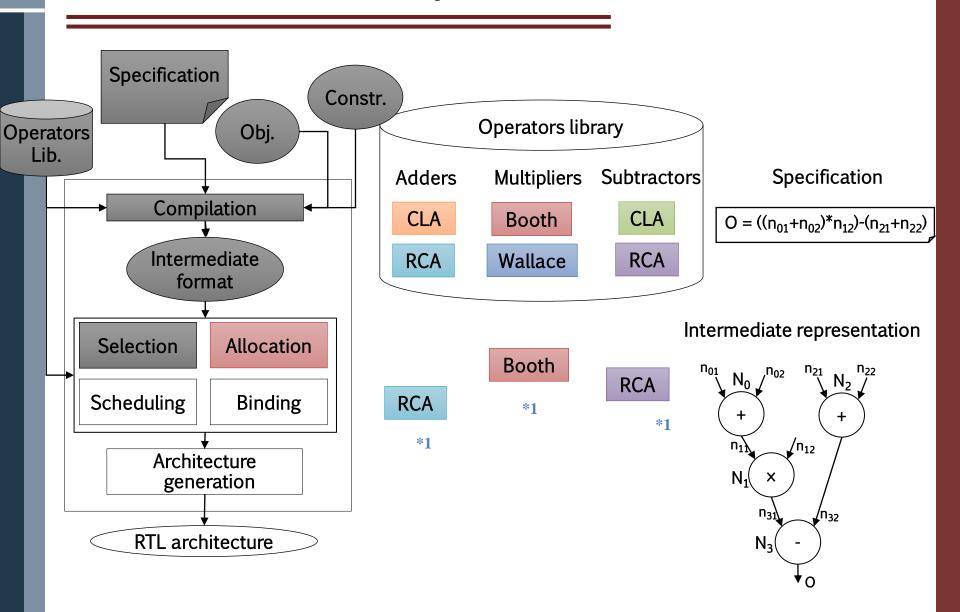


HLS steps: Selection



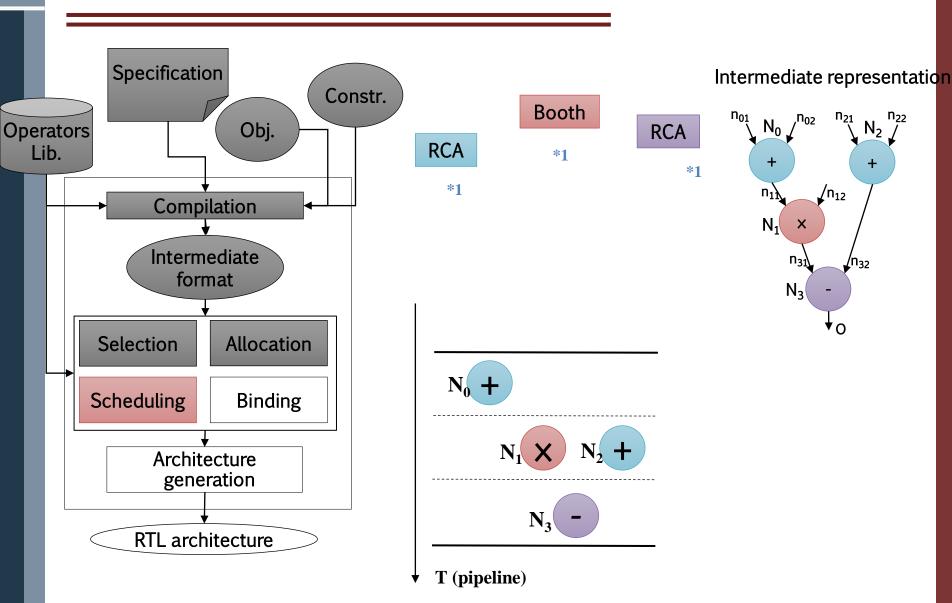


HLS steps: Allocation



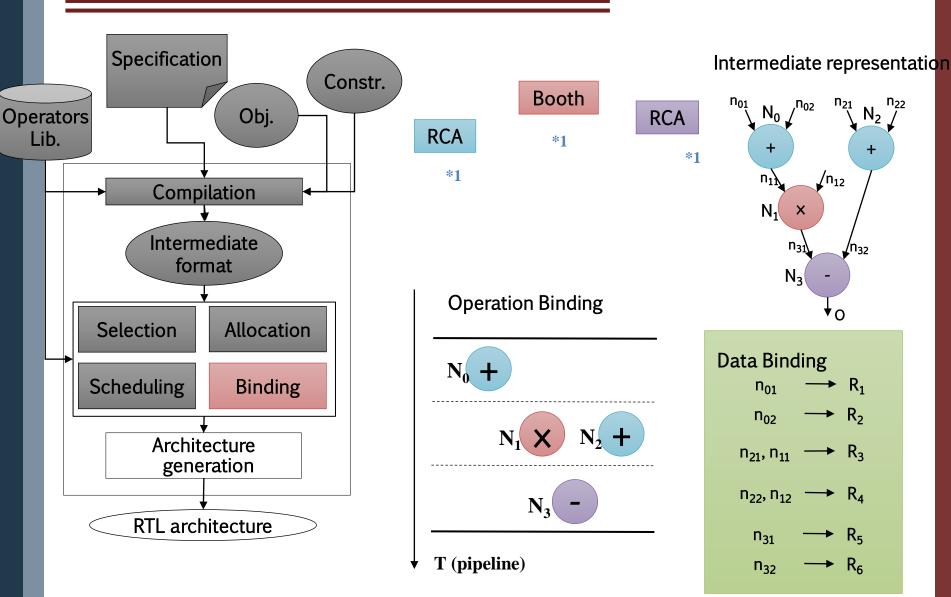


HLS steps: Scheduling



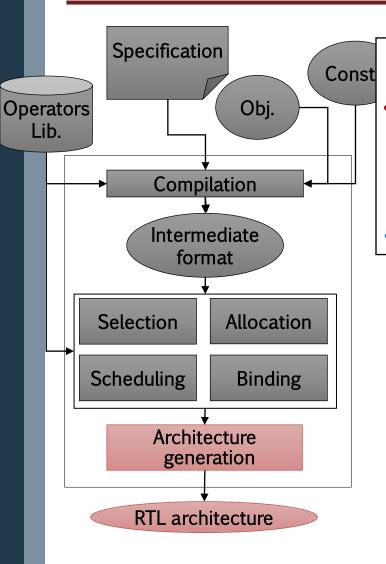


HLS steps: Binding





HLS steps: Binding

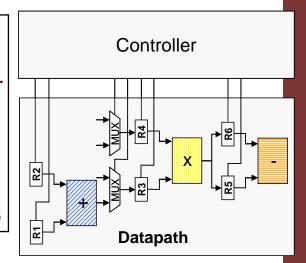


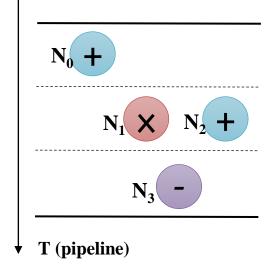
Controller

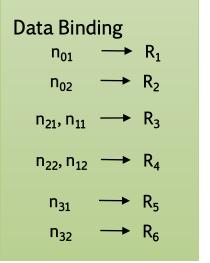
- FSM controller
- Programmable controller

Datapath components

- Storage components
 - Functional units
- Connection components









Xilinx's Vivado SDK

The FPGA development tool

- > Starting from C or RTL...
- > ...generates and deploys the IP on the FPGA
- > ..as well as SW artifacts to interact with them (drivers)

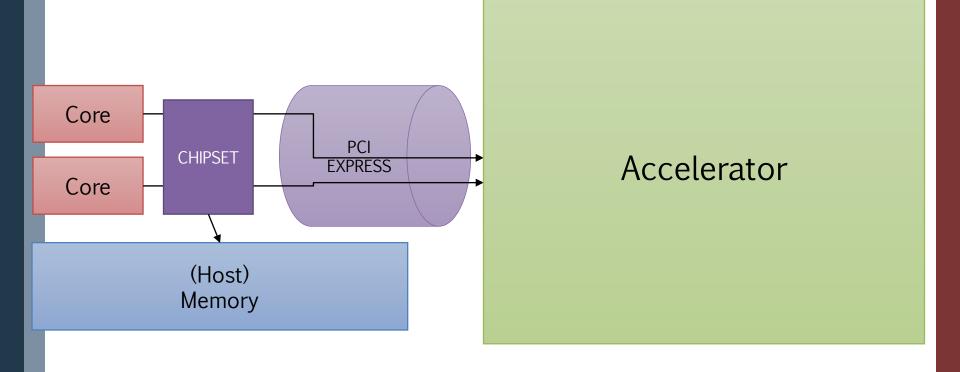
> Let's see it in action!

Heterogeneous systems



Host-accelerator model

- > Multi-core General purpose host
 - The "traditional" core
- > Coupled with a co-processor/accelerator

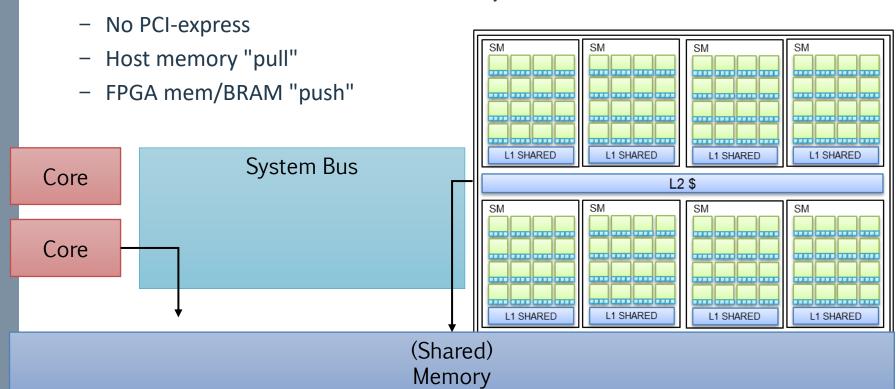




Do you remember (i) GPGPUs?

Integrated GP-GPUs for embedded platforms

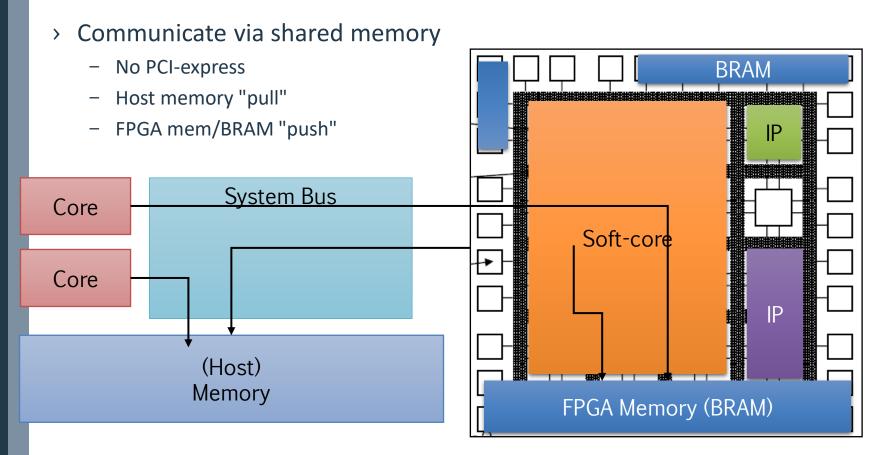
- > Host + accelerator model
- > Communicate via shared memory





FPGA-based accelerators

- > Can create hundreds of small HW accelerators (de/crypt, de/coders)
- Can even create a single core (as co-processor)
 - Soft-cores





Example: Xilinx Zyng L3 Memory **Dual-core ARM host Processing System** Programmable Logic **Programmable** Static Memory Controller **Dynamic Memory Controller** Quad-SPI, NAND, NOR DDR3, DDR2, LPDDR2 Shared memory Logic: System Gates, Host caches AMBA® Switches 2x SPI AMBA® Switches DSP, RAM 2x 12C Host SPM (On-chip Memc ARM® CoreSight™ Multi-core and Trace Debug 2x CAN External DDR banks NEON™/FPU Engine NEON™/FPU Engine 2x Uart Cortex -A9 MPCore 32/32 KB I/D Caches Cortex ""-AY MPCore Connectivity 32/32 KB I/D Caches ACP 512 KB L2 Cache **Snoop Control Unit (SCU)** CP!O AMBA AXI connector 256 KB On-Chip Memory **Timer Counters** 2x SDIO with DMA Highly scalable General Interrupt Controller Configuration 2x USB with DMA Cache coherency port **ACP**

XADC

Multi Standards I/Os (3.3V & High Speed 1.8V)

L1/2 Memory **PCle**

Multi Gigabit Transceivers

Xilinx FPGA SoCs

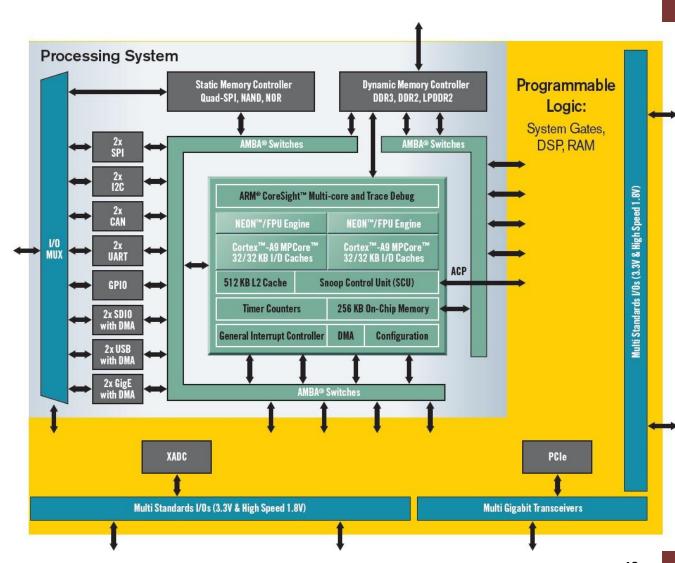


Xilinx Zynq-7000

Dual-core ARM host

Programmable Logic

Shared memory



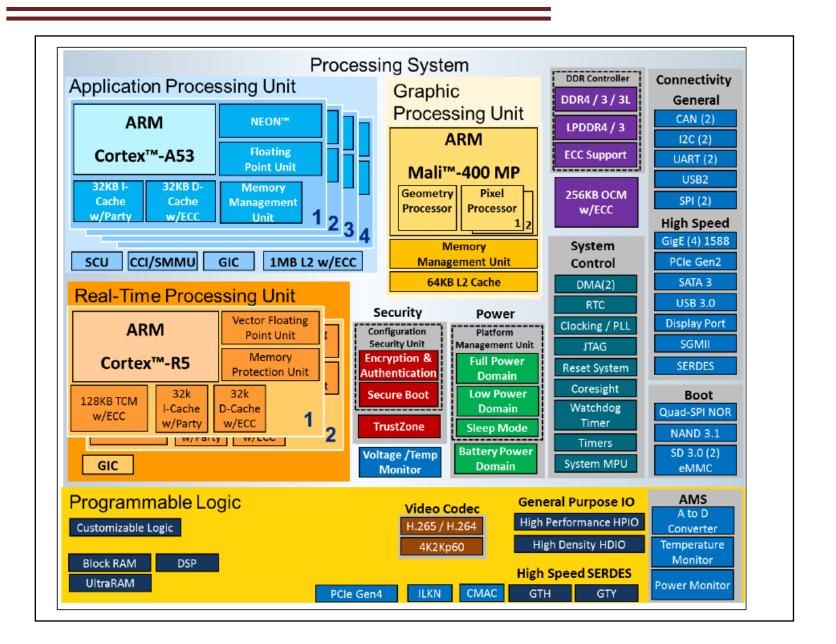


Xilinx Zynq-7000

FAMILY	PART	Logic Cells (K)	Block RAM (Mb)	DSP Slices	Maximum I/O Pins	Maximum Transceiver (Video Code Unit (VCU)
ZYNQ-7000							
	Z-7010	28	2,1	80	100	-	-
	Z-7015	74	3,3	160	150	4	-
	Z-7020	85	4,9	220	200	-	-
	Z-7030	125	9,3	400	250	4	-
	Z-7035	275	17,6	900	362	16	-
	Z-7045	350	19,1	900	362	16	-
	Z-7100	444	26,5	2020	400	16	-



Xilinx Zynq Ultrascale+





Xilinx Zynq Ultrascale portfolio

> Zynq UltraScale+ CG

- Dual-core Cortex-A53 and a dual-core Cortex-R5 real-time processor
- Programmable logic
- Optimized for industrial motor control, sensor fusion, and industrial IoT applications

> Zynq UltraScale+ EG

- Quad-core Cortex-A53 and dual-core Cortex-R5 real-time processors
- Mali-400 MP2 graphics processing unit + programmable logic
- Next-generation wired and 5G wireless infrastructure, cloud computing, and Aerospace and Defense applications

> Zynq UltraScale+ EV

- EG platform + integrated H.264 / H.265 video codec
- Multimedia, automotive ADAS, surveillance, and other embedded vision applications



Xilinx Zynq Ultrascale+

FANAULY	DADT	Lasia Calla (K)	Disale DANA (NAIs)	DCD Clines	Massimos I/O Dina	NASSISSES TOPOGRAPION	() (; d = = C = d = = ; + () (C)
FAMILY	PART	Logic Cells (K)	Block RAM (Mb)	DSP Slices	Maximum I/O Pins	iviaximum transceiver	Video Code Unit (VCU)
7VNO Liltra Caala L CC							
ZYNQ UltraScale+ CG	ZU2CG	102	ГЭ	240	252		
		103					-
	ZU3CG	154					-
	ZU4CG	192					-
	ZU5CG	256					-
	ZU6CG	469					-
	ZU7CG	504					-
	ZU9CG	600	32,1	2520	328	-	-
ZYNQ UltraScale+ EG							
	ZU2EG	103					-
	ZU3EG	154					-
	ZU4EG	192					-
	ZU5EG	256	23,1	1248	252	-	-
	ZU6EG	469	25,1	1973	328	-	-
	ZU7EG	504	38	1728	464	-	-
	ZU9EG	600	32,1	2520	328	-	-
	ZU11EG	653	43,6	2928	512	-	-
	ZU15EG	747	57,7	3528	328	-	-
	ZU17EG	926			668	-	-
	ZU19EG	1143	70,6	1968	668	-	-
ZYNQ UltraScale+ EV							
	ZU4EV	192	18,5	728	252	-	1
	ZU5EV	256			252	-	1
	ZU7EV	504					1



Zedboard

- Complete development kit with Xilinx Zynq-7000 SoC
- Basic support for rapid prototyping and proof-of-concept development
- > Small ©





UltraZed

- > System-On-Module (SOM)
- > Based on the Ultrascale architecture: no host!
- > Packages system memory, Ethernet, USB, and configuration memory needed for an embedded processing system

- > UltraZed EG
- > Ultrazed EV



Xilinx Pynq: Python for Zynq

> Open-source project from Xilinx for design

> Uses Python language and libraries

Maximizes productivity

Processor: Dual-Core ARM® Cortex®-A9

FPGA: 1.3 M reconfigurable gates

Memory: 512MB DDR3 / FLASH

Storage: Micro SD card slot

Video: HDMI In and HDMI Out

Audio: Mic in, Line Out

Network: 10/100/1000 Ethernet

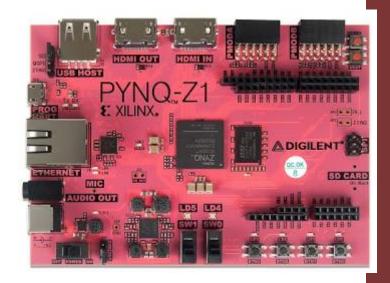
Expansion: USB Host connected to ARM PS

Interfaces: 1x Arduino Header, 2x Pmod (49 GPIO)

GPIO: 16 GPIO (65 in total with Arduino and Pmods)

Other I/O: 6x User LEDs, 4x Pushbuttons, 2x Switches

Dimensions: 3.44" x 4.81" (87mm x 122mm)



Programming heterogeneous systems

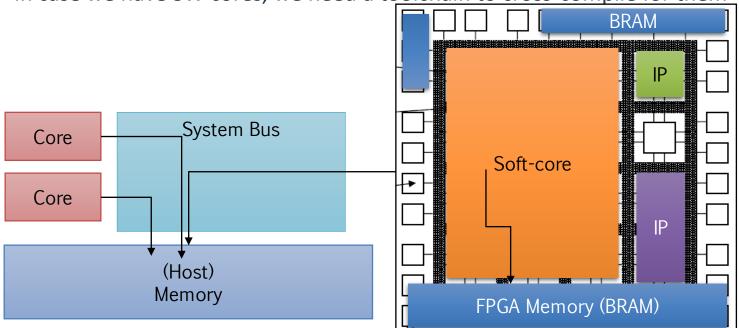


Heterogeneous programming

Besides a tool to generate the actual IPs, we need

- A way to efficiently offload (pre-compiled) bitcode on the FPGA
 - On-the-fly Dynamic Partial Rreconfiguration (DPR)
- > Simple offloading subroutines to the newly created HW blocks
 - To increase productivity

In case we have SW cores, we need a toolchain to cross-compile for them





1) custom/"by hand"/CAD

Code generated by logic synthesis tool

- > Step 1 generate the bitcode of the accelerator
 - Vivado HLS
- > Step 2 plug the accelerator in a design
 - Vivado
 - Include processing system (ARM host) + accelerator + IC + ...
- > Step 3 generate the design
 - Bitcode ready to be installed of the IP
 - Architecture configuration files (memory maps...)
 - Software for host + drivers to communicate with the IP

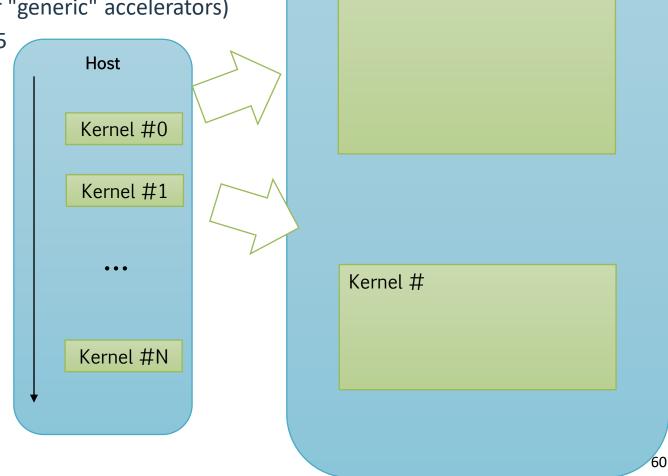
Let's see this in action!



2) offload-based programming

> Offload-based programming model

- CUDA (for NVIDIA GPUs)
- OpenCL (for "generic" accelerators)
- OpenMP 4.5



Device

Kernel #0



OpenCL

- OpenCL was initiated by Apple and maintained by the Khronos Group (also home of OpenGL) as an industry standard API
 - For cross-platform parallel programming in CPUs, GPUs, DSPs, FPGAs,...
- > OpenCL host code is much more complex and tedious due to desire to maximize portability and to minimize burden on vendors



OpenCL program

An OpenCL "program" is a C program that contains one or more "kernels" and any supporting routines that run on a target device

An OpenCL kernel is the basic unit of parallel code that can be executed on a target (OpenCL Program

Kernel C

> In our case, an FPGA

Misc support functions

Kernel A

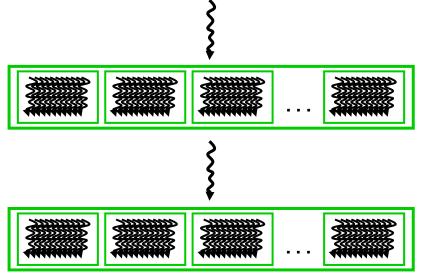
Kernel B



OpenCL execution model

- > Integrated host+device app C program
 - Serial or modestly parallel parts in host C code
 - Highly parallel parts in device SPMD kernel C code

> Queues of command/data transfer to be executed on the device





OpenCL kernels – software version

- Code that executes on target devices
- > Kernel body is instantiated N times (data parallel) work items
- > Each OpenCL work item gets a unique index

> In the FPGA case we use IP drivers instead of this



Host code – create exec ctx

```
cl_int clerr = CL_SUCCESS;
cl_context clctx = clCreateContextFromType(0, CL_DEVICE_TYPE_ALL, NULL, NULL, &clerr);
size_t parmsz;
clerr = clGetContextInfo(clctx, CL_CONTEXT_DEVICES, 0, NULL, &parmsz);
cl_device_id* cldevs = (cl_device_id *) malloc(parmsz);
clerr = clGetContextInfo(clctx, CL_CONTEXT_DEVICES, parmsz, cldevs, NULL);
cl_command_queue clcmdq = clCreateCommandQueue(clctx, cldevs[0], 0, &clerr);
```



Host code – create data buffers



Host code – device config setting

```
clkern=clCreateKernel(clpgm, "vadd", NULL);
...

clerr= clSetKernelArg(clkern, 0, sizeof(cl_mem), (void *)&d_A);

clerr= clSetKernelArg(clkern, 1, sizeof(cl_mem), (void *)&d_B);

clerr= clSetKernelArg(clkern, 2, sizeof(cl_mem), (void *)&d_C);

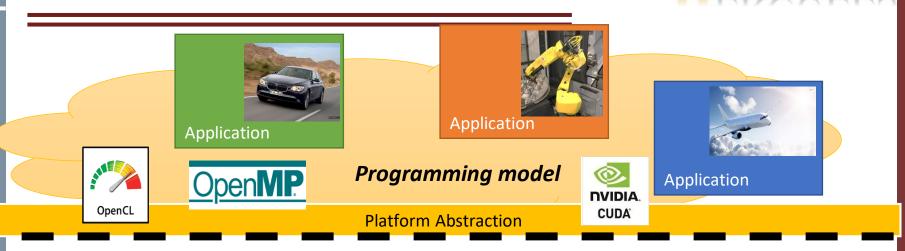
clerr= clSetKernelArg(clkern, 3, sizeof(int), &N);
```

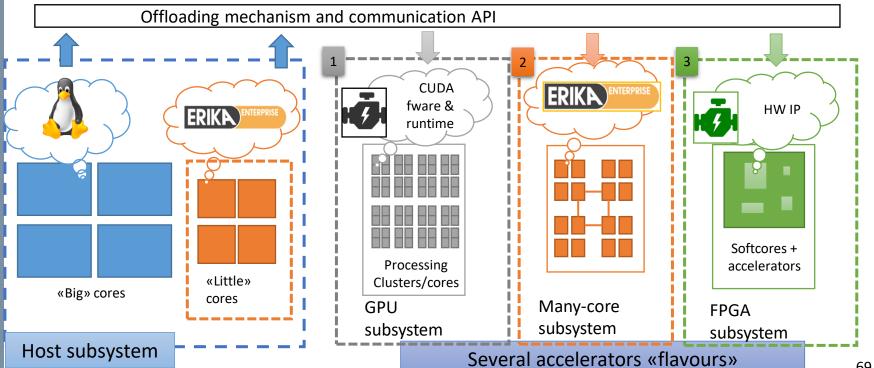


Host code – kernel launch



The Hercules framework HERCULES



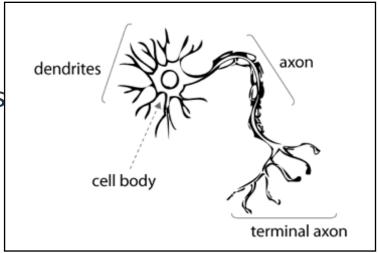


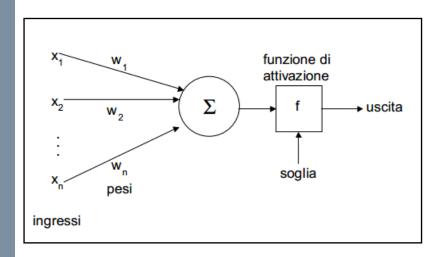
Neural Networks on FPGA accelerators



Neural networks

- > Bio-inspired
- > Based on neurons arranged in layers
 - And sub-layers
- > Convolutional neural network
 - Neurons perform Convolutions



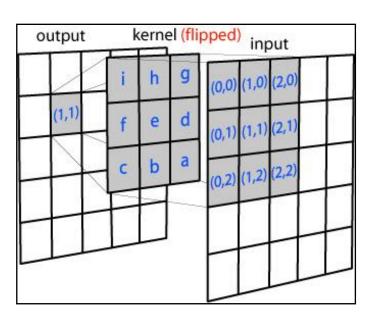




Convolution

- > Computation-intensive
- > Suitable for implementation in hardware
- > In computer vision, blurring

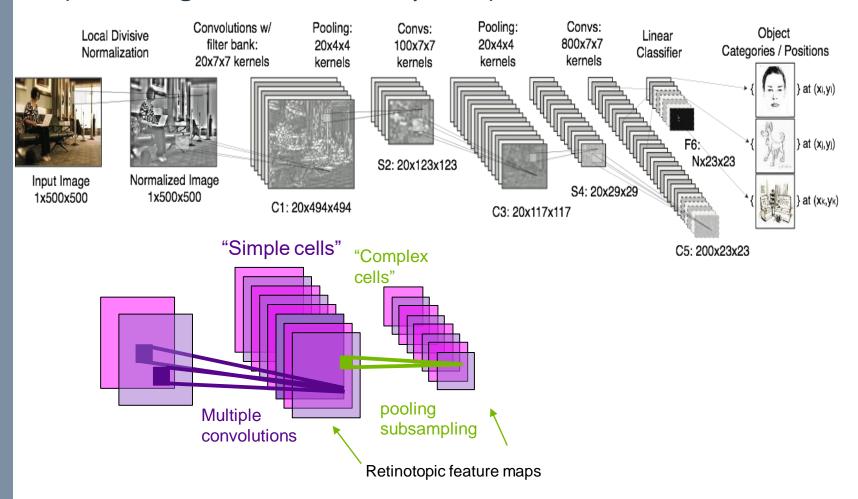
$$(fst g)(t) \stackrel{ ext{def}}{=} \int_{-\infty}^{\infty} f(au)g(t- au)\,d au \ = \int_{-\infty}^{\infty} f(t- au)g(au)\,d au.$$





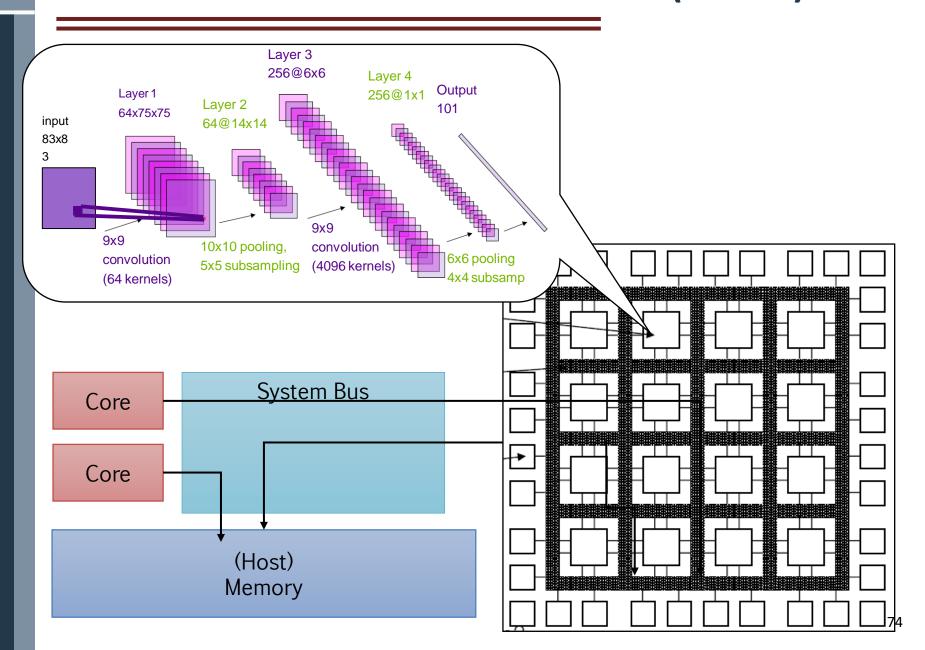
The convolutional net model

> (Multistage Hubel-Wiesel system)





The convolutional net model (cont'd)





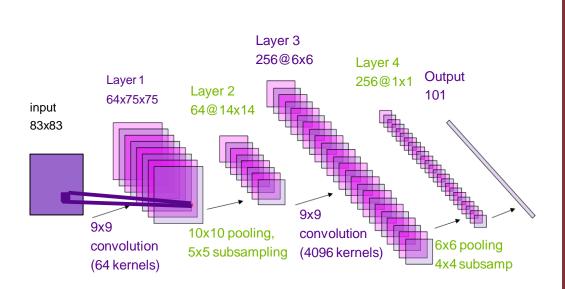
Network parameters

Network topology

- > How many layers and sublayers?
- > How big they are?
- > How are they connected?

Neuron type

- > CNN
- int/float datatypes
- > How to perform pooling?





Training a network

"The training problem"

- > To set the weights/CNN kernels
- > Training set must be huge

A "big data" problem

- > Why do you think Google does self-driving cars?
- > Why do you think big cloud players want our data?





CNN on FPGA

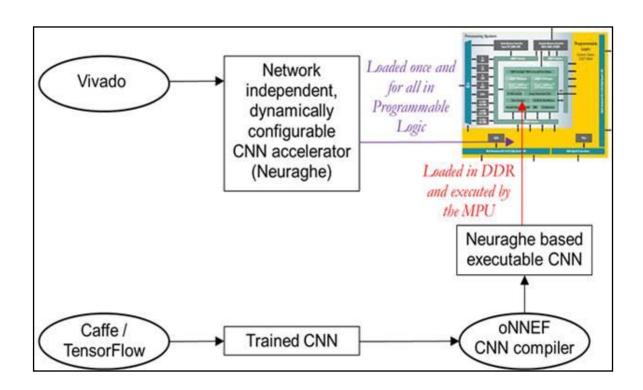
- > We can implement one or more complete CNN layers on FPGA
 - How many?
- > We can use float, int, datatypes
 - Int are smaller, but still efficient
- > Binaries NN, where input weights are +-1
 - Smaller, more sutable for area-constrained

Name	Supported Architecture
ZynqNet	XC-7Z045
BNN	
GraphGen-based CNN	Z-7020
Pre-Trained CNN based on LeNet5	
GoogLeNet	
AlexNet	
VGG-16	ZCU102
SSD-300	
FCN-AlexNet	



Nuraghe NN on FPGA

- > Trained with ML frameworks Caffè and TensorFlow
- > Nuraghe accelerator, configurable for specific CNN
- CNN compiler, which translates the CNN description from Caffè or TF in a program which runs on Nuraghe/Zynq





References



Course website

http://hipert.unimore.it/people/paolob/pub/Industrial Informatics/index.html

My contacts

- > paolo.burgio@unimore.it
- http://hipert.mat.unimore.it/people/paolob/

Resources

- > Xilinx Zynq-7000 All Programmable SoC => https://www.xilinx.com/products/silicon-devices/soc/zynq-7000.html
- > Pynq => http://www.pynq.io/
- > Xilinx Ultrascale => https://www.xilinx.com/products/technology/ultrascale.html
- A "small blog« => http://www.google.com