

# Αρχιτεκτονική Προηγμένων Υπολογιστών και Επιταχυντών Lab 1 Report

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# Chapter 1

## Exercise 1

### 1.1 Core concepts

- LUTs: **Look Up Tables** are programmable truth tables inside an FPGA that implements logic operations.
- DSPs: **Digital Signal Processing** units are specialized hardware whose purpose is to do mathematical operations (mainly multiplication and division) really fast.
- BRAM: **Block Random Access Memory** is internal temporary memory smaller but faster than DRAM.
- DRAM: **Dynamic Random Access Memory** is external temporary memory slower but bigger than BRAM.
- FFs: **Flip-Flops** are simple elements that store 1 bit.

### 1.2 Core concepts applied on our code

In our implementation we create two instances of the A, B and C 2D arrays. One which holds the data stored inside DRAM and one which holds the data inside BRAM. The reason why we decided to split the data is so that we could implement array\_partition later on, which needs our matrices to reside in BRAM. When it comes to FFs and LUTs they are directly proportional to the size of our matrices because enlarging the matrices' dimensions leads to more hardware needed to translate the software. Additionally, it turned out that since our math operations are simple subtractions between low bit unsigned integers, DSPs were not utilized.

### 1.3 Interface

AXI is a family of AMBA (Advanced Microcontroller Bus Architecture) protocols.

#### 1.3.1 m\_axi

A full AXI-4 master interface which enables reading and writing A, B, C directly to DRAM.

```
#pragma HLS INTERFACE m_axi port=A depth=1024 offset=slave  
#pragma HLS INTERFACE m_axi port=B depth=1024 offset=slave  
#pragma HLS INTERFACE m_axi port=C depth=1024 offset=slave
```

- **m\_axi**:

- **offset=slave**: the runtime base address is provided via an AXI-Lite register
- **depth=1024**: number of elements in the array

### 1.3.2 s\_axilite

An AXI4-Lite slave interface that the CPU will use to program the accelerator. Used for:

- Writing A's addresses
- Writing B's addresses
- Writing C's addresses
- Starting the Kernel

```
#pragma HLS INTERFACE s_axilite port=A bundle=control
#pragma HLS INTERFACE s_axilite port=B bundle=control
#pragma HLS INTERFACE s_axilite port=C bundle=control
#pragma HLS INTERFACE s_axilite port=return bundle=control
```

## 1.4 Pragmas

### 1.4.1 bind\_storage

We bind our local variants of the A, B, C matrices as bram using 2 ports.

```
#pragma HLS bind_storage variable=A_local type=ram_2p impl=bram
#pragma HLS bind_storage variable=B_local type=ram_2p impl=bram
#pragma HLS bind_storage variable=C_local type=ram_2p impl=bram
```

### 1.4.2 array\_partition

We partition the second dimension of our matrices so that we can increase our data throughput. Since we use cyclic partitioning the data is split like so:

Bank 1	Bank 2	Bank 3	Bank 4
1st element	2nd element	3rd element	4th element
5th element	6th element	7th element	8th element
...	...	...	...

This permits us to fetch  $x$  elements per cycle where  $x$  the number of banks. Here follows

```
#pragma HLS array_partition variable=A_local cyclic factor=8 dim=2
#pragma HLS array_partition variable=B_local cyclic factor=8 dim=2
#pragma HLS array_partition variable=C_local cyclic factor=8 dim=2
```

### 1.4.3 unroll

Loop unrolling takes a number of individual loops from a for, depending on the unroll factor, and effectively stacks them together allowing their parallel execution. When a for loop is not unrolled the next loop's operations can't begin, and thus can't be parallelized, since we are not yet sure of the branch destination.

Since, as we were taught, pipelining flattens the loops automatically, we avoided using loop unrolling but this is the notation we used on our experiments:

```
#pragma HLS unroll factor=4
```

#### 1.4.4 pipeline

Pipelining allows identical assembly operations to be parallelized by executing them with 1 clock cycle time delay. This is possible due to the fact that we utilize different hardware on each of an operation's cycles. Expanding on that notion we can also pipeline different operations as long as they don't utilize the same hardware at the same cycle. Here we attempt to pipeline our code every 1 cycle:

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

# Chapter 2

## Exercise 2

Name/Loop	Latency (cycles)	Latency (ns)	Interval	Pipelined	BRAM	DSP	FF	LUT	URAM
Top function	262295	2.623E6	262296	no	12	0	46751	12036	0
Loop 1	131145	1.311E6	131145	no	0	0	40567	1764	0
Loop 2	65538	6.550E5	65538	no	0	0	55	255	0
Loop 3	65539	6.550E5	65539	no	0	0	1061	1055	0

For an array of  $256 \times 256$  we get the following results:

Estimated clock period	7.300ns
Worst case latency	262295 cycles
Number of DSP48E used	0
Number of BRAMs used	12
Number of FFs used	46751
Number of LUTs used	12036

# Chapter 3

## Exercise 3

Here are the results obtained through the cosimulation:

Total Execution Time	2,675,185.0 ns
Min Latency	267446 cycles
Avg Latency	267446 cycles
Max Latency	267446 cycles

# Chapter 4

## Exercise 4

### 4.1 Part 1

Compared to the previous two exercises where we didn't use any non interface pragmas, here we will integrate some hls directives for optimization with regard to latency. On one hand we use *array\_partition* to create blocks of memory each with it's own port:

```
#pragma HLS array_partition variable=A_local complete dim=2
#pragma HLS array_partition variable=B_local complete dim=2
#pragma HLS array_partition variable=C_local complete dim=2
```

On the other hand we use pipeline pragmas in every for loop for parallelization:

```
for (int i = 0; i < HEIGHT; ++i)
    #pragma HLS pipeline II=1
    for (int j = 0; j < WIDTH; ++j)
```

These optimizations achieve in our array of  $256 \times 256$  a Latency of **3840 cycles**. In other words 75 times faster latency than originally.

Experimentation with the second dimension of the Array (WIDTH):

WIDTH	Latency (cycles)	Execution Time	BRAM	LUTS
64	1175	30,175ns (2945cyc)		19488
128	1944	55,735ns (5501cyc)	384	29889
256	3480	106,945ns (10622cyc)	768	50600
512	6552	209,365ns (20864cyc)	1536	92090

### 4.2 Part 2

The optimal, between our tested configurations, setup was a pure hardware translation of the circuit using lots of LUTs and FFs while omitting BRAM all-together. Below are its attributes:

Parameter	Value
Estimated Clock period	3.934 ns
Number of DSPs	0
Number of BRAMs	0

Parameter	Value
Number od FFs	67644
Number of LUTs	99181
Total Execution Time	0
Min Latency	258 cycles
Avg Latency	258 cycles
Max Latency	258 cycles

### 4.3 Part 3

The total acceleration of our optimal solution is:

$$\begin{array}{c} \hline \text{Acceleration} \\ \hline \hline 1886.509\% \\ \hline \end{array}$$

# Chapter 5

## Zip Contents

The zip we attach for grading contains:

1. Two Testbench files each one using a different validation method
  - random number generation or
  - standard pattern
2. Two Source files
  - One optimal and
  - the one used in most of our tests
3. This report