

Project: CNN Acceleration System

EE470 – System-on-Chip (SoC) Design

Kyung Hee University
Electrical Engineering

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Seungkyu Choi (seungkc@khu.ac.kr)

Project Overview

- **Goal:** Build an end-to-end CNN inference system on Zynq-7000
- **Model:** LeNet-1 (modified) on MNIST (28×28 grayscale)
- **Platform:** Arty Z7-20
- **Tools:** Vivado + Vitis (2024.2)

- **Implementation Scope:**

Students will design the PL accelerator, develop a PS-side application, and integrate them through an AXI-based interface.

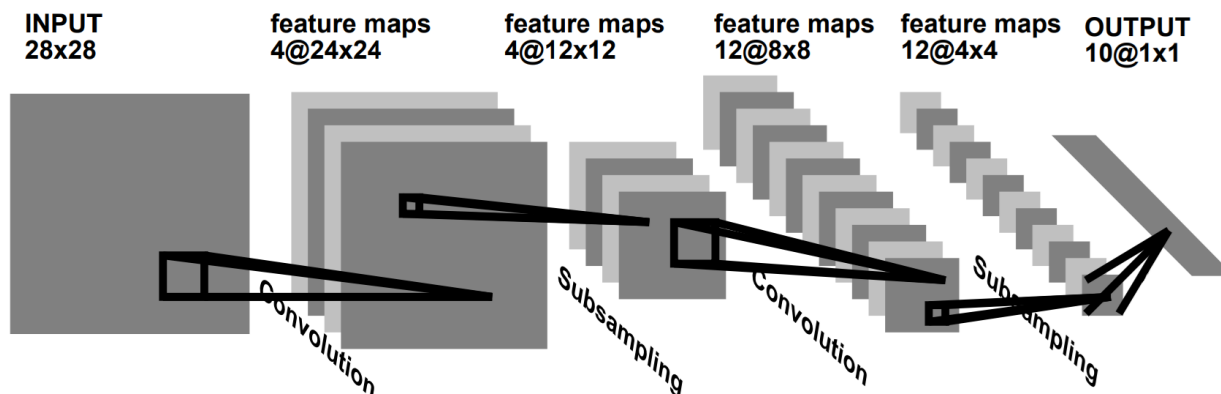
- **Focus Areas:**

BRAM-based buffering, tiling, and sustained dataflow for efficient computation.

(Target Metrics: Latency / Accuracy / Hardware Utilization)

CNN Model

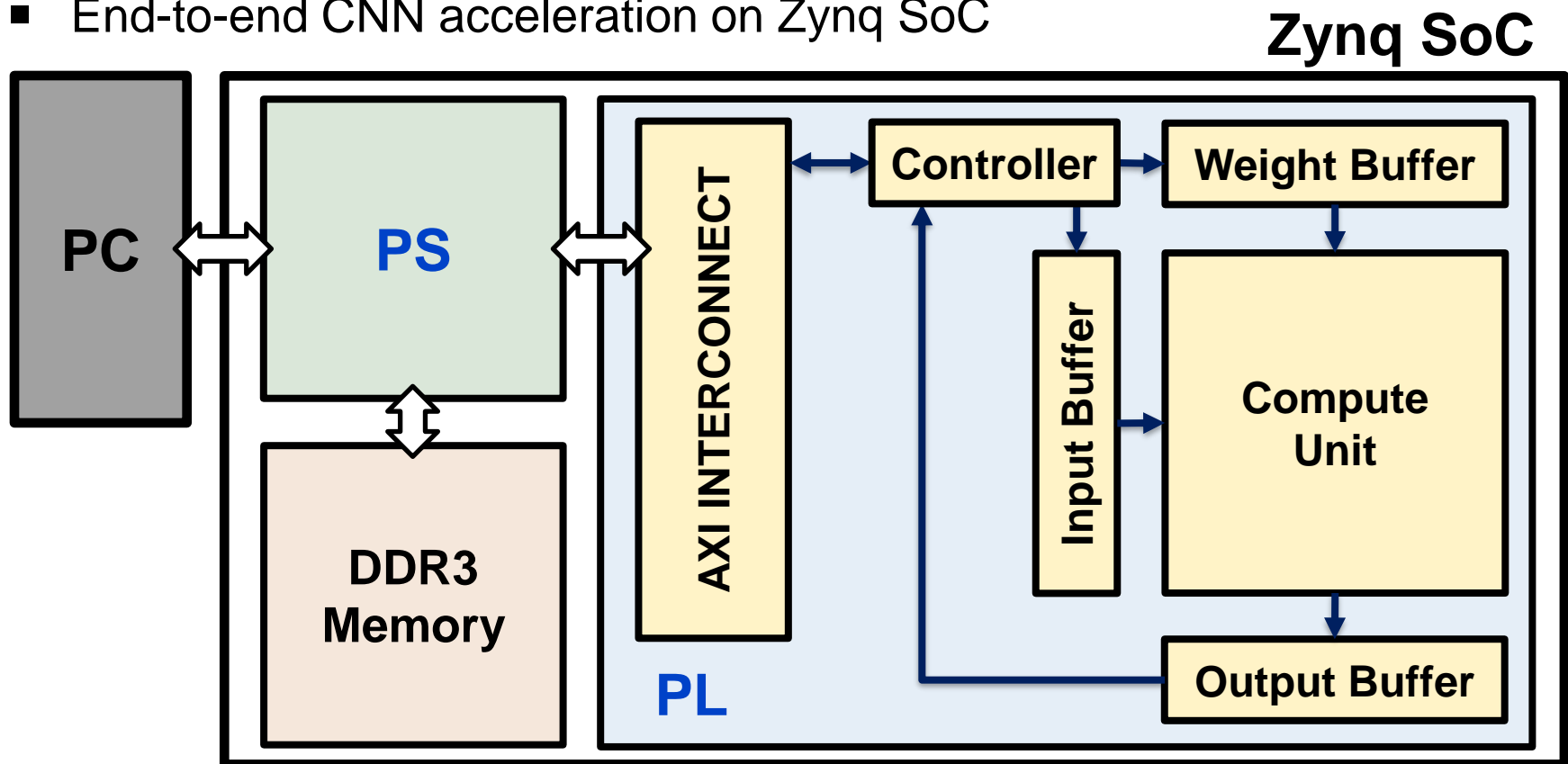
- LeNet-1 (modified) on MNIST(28×28 grayscale, uint8)



- **CONV1** : 4×(5×5) kernel, stride=1, no padding or bias → **Output**: 4×24×24
- **POOL1** : 2×2 max-pool, stride=2 → **Output**: 4×12×12
- **CONV2** : 12×(5×5) kernel, stride=1, no padding or bias → **Output**: 12×8×8
- **POOL2** : 2×2 max-pool, stride=2 → **Output**: 12×4×4
- **FC** : Flatten (12×4×4=192), no bias → **Output**: 10 classes
- ReLU is used as an activation function after each layer.
- Quantized Weights(uint8) are provided in .h format for each layer.(CONV1, CONV2, FC)

System Architecture

- End-to-end CNN acceleration on Zynq SoC



- PS (ARM Cortex-A9): Control, data management
- PL (FPGA fabric): CNN computation (CONV/FC)
- Communication: AXI4-Lite

Design Requirements

- Model size must remain unchanged.
(Keep the same given LeNet-1 structure and feature map dimensions.)
- CONV & FC must execute in PL
- Implement BRAM-backed buffering
- Develop a PS-side Vitis application(main.c) to control and test the CNN accelerator.
(You are only allowed to add code in the permitted sections.)
- Complete Vivado synthesis/implementation and ensure timing closure.
(Check WNS/TNS.)

Flexibility & Advanced Options

■ Compression techniques & Optimizations

- Quantization
- Pruning
- Word Alignment
- Weight reordering (e.g., transpose, im2col)
- Finetune **(Include before/after accuracy & training log in your report)*

➡ You may **customize** your own weight file.

(Describe the change in file size and your optimization method in the report)

- **Interface:** You may use AXI-DMA/Stream beyond just AXI4-Lite.
- **Compute Unit:** The internal compute architecture is open to your design.
(e.g., systolic array, parallel MAC array)

Evaluation Criteria

■ Dataset:

- From the MNIST test set (10,000 images), a 1,000-image subset will be provided for development and debugging.
- The grading subset (another 1,000-image subset) will **not** be provided.

■ Evaluation Metric:

- **Latency:** End-to-end latency measured from the first PS-PL transaction to result return.
- **Accuracy:** Computed on PS using logits from the PL FC output.
Report Top-1 accuracy (%) on the test set.
- **Hardware Utilization:** LUT / FF / BRAM / DSP usage
(post-implementation report)

■ Note:

Overall trade-offs among latency, accuracy, and resource efficiency will be well considered.

Design Workflow

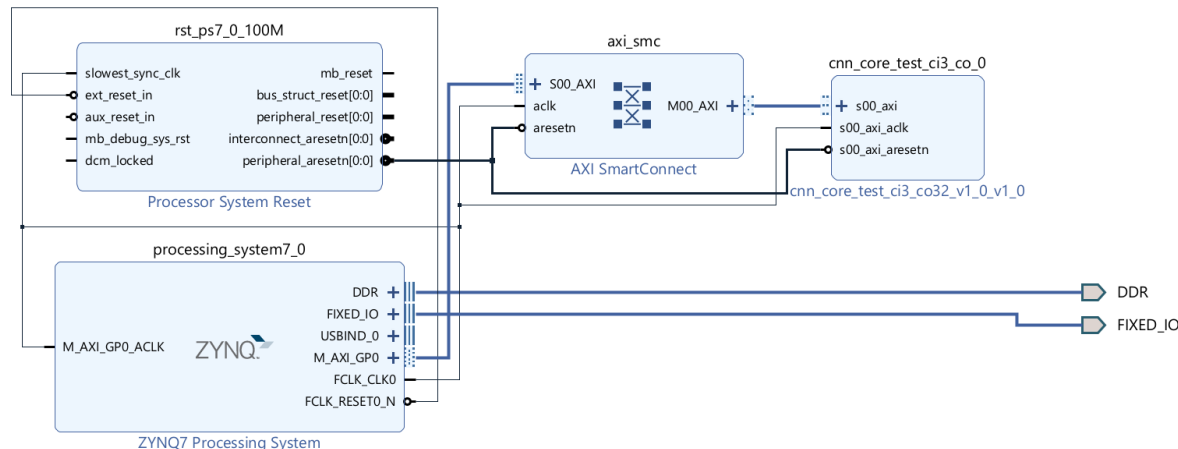
< Vivado: Hardware Design >

1. Create IPs from Verilog RTL modules.

- Develop your CNN accelerator modules (e.g., CONV, FC, Controller).
- Package them as custom IPs.

2. Build a System Block Design

- Instantiate your custom IP along with PS, BRAM, AXI Interconnect, etc.
- Configure address mapping and interface connections.



Design Workflow

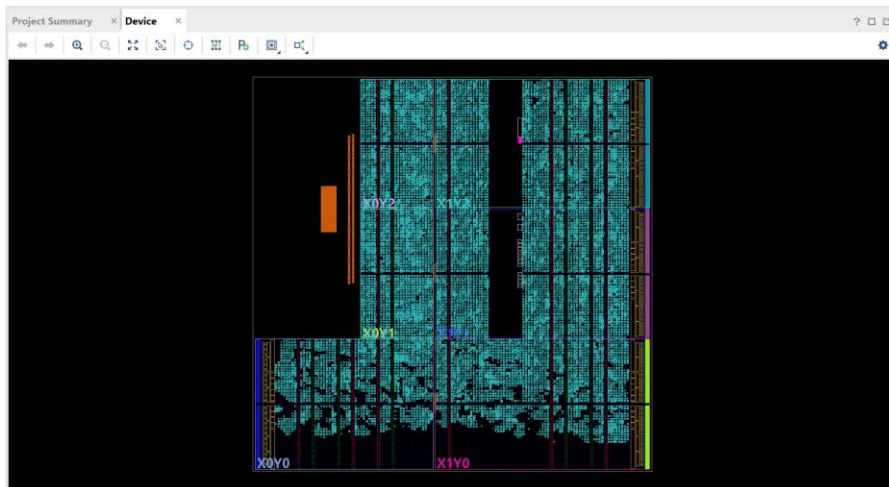
< Vivado: Hardware Design >

3. Generate Bitstream

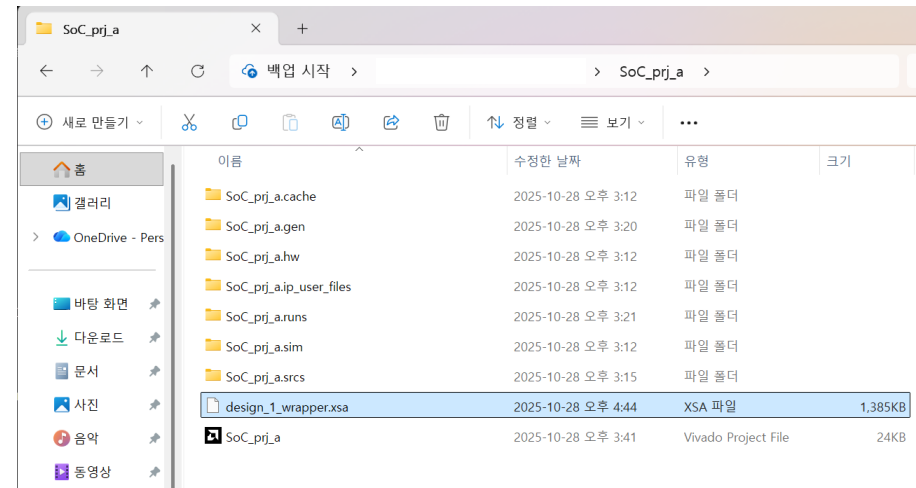
- Run synthesis and implementation.
- Check **timing, power, and utilization reports**. (WNS/TNS, LUT/FF/DSP/BRAM)

4. Export Hardware (.xsa)

- Create a top-level wrapper and export the design as **.xsa** for Vitis.



[Device View]



[.xsa file]

Design Workflow

< Vitis: Software Integration & Testing >

1. Import Hardware Platform (.xsa)

- Create a new workspace with the exported hardware design as the base platform.

2. Develop Application Project

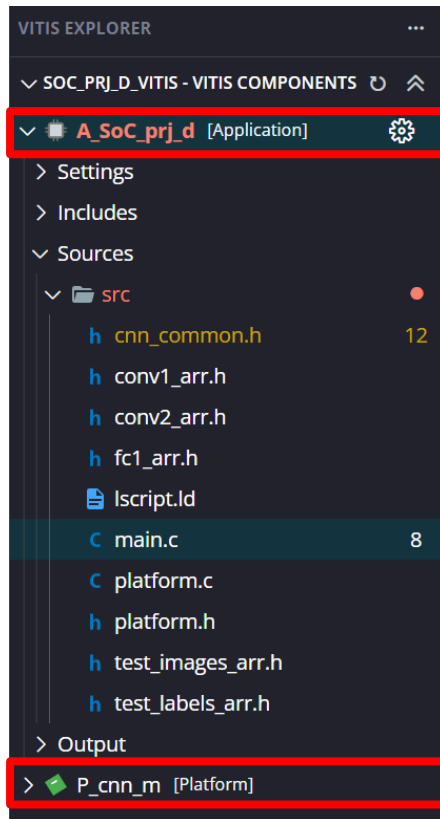
- Use the provided `main.c` skeleton to control your PL accelerator via AXI.
- Measure accuracy and latency, and compare them with the software Implementation on PS.

3. Run on Hardware (Serial Output)

- Connect via UART/Serial Port to monitor execution.
- Display classification accuracy and total inference latency on terminal.

Design Workflow

< Vitis: Software Integration & Testing >



[Vitis Components]

```
Opened with baud rate: 115200
***** SoC CNN Acceleration System *****
Press '1' Start Test
Press '2' to exit
Selection:1

[PS vs PL E2E CNN] N_TEST=1000
>>> CNN Running in PS...
>>> Loading Weights on PL...
>>> CNN Running in PL...

=== Summary ===
PS  Acc = 986/1000 = 98.60% | Total = 8665773.53 us | Avg/img = 8665.77 us
PL  Acc = 986/1000 = 98.60% | Weight upload = 240.80 us
PL  Inference only: Total = 1018604.20 us | Avg/img = 1018.60 us
PL  Cold-start total (upload + infer) = 1018845.00 us | Avg/img (including upload) = 1018.85 us
Speedup (PS/PL, steady-state avg) = 8.51x
```

[Test Result]

**Shown for illustration only
(NOT actual result)*

Fine-tuning Guideline

■ MNIST Dataset Access for Fine-tuning (Option)

- **MNIST:** A dataset of 28x28 grayscale images of handwritten digits (0-9)
- **How to Download**

```
from torch vision import datasets
import torchvision.transforms

train_ds = datasets.MNIST(
    root="./my_root",
    train=True,
    transform= ~~ ,
    download=~~)
```

- **Root:** where to store MNIST locally
- **Train:** True = training set, False = test set
- **Transform:** preprocess per image (e.g. ToTensor(), PILToTensor())
- **Download:** download if not already present
- ❖ **Reference:** MNIST docs
(<https://docs.pytorch.org/vision/main/generated/torchvision.datasets.MNIST.html#torchvision.datasets.MNIST>)

Submission Rule (Due: Dec 14 23:59)

- Submission file: Zip File *File Name: prj_xx.zip (xx = team number 01 ~ N)*
 - Complete SoC Project Package
 - Entire Vivado project folder (including .xpr)
 - Vitis main.c file
 - Weight files (.h) *(if modified)*
 - PPT & Report (.pdf)
 - Should describe overall system architecture, implementation details, and control flow.
 - Presentation time: 10 - 12 minutes.
 - Report length: under 10 pages. *(Korean allowed)*