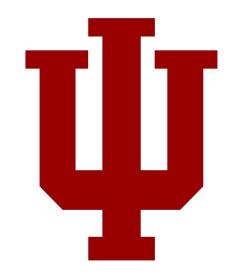
# 15: Hardware Pipelining II

ENGR 315: Hardware/Software CoDesign

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Some material taken from: <a href="https://github.com/trekhleb/homemade-machine-learning/tree/master/homemade/neural\_network-1/">https://github.com/trekhleb/homemade-machine-learning/tree/master/homemade/neural\_network-http://cs231n.github.io/neural-networks-1/</a>



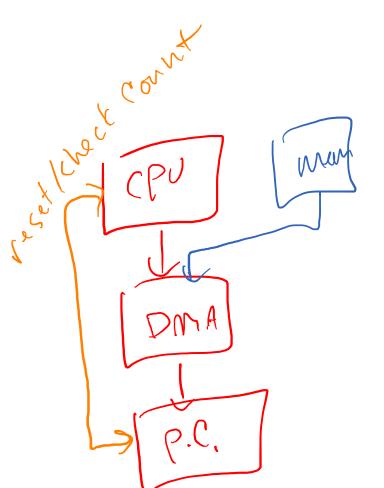
## Announcements

• P

• P6 is out (I think)

• Exam in 3 weeks

# P5: Adds DMA + AXI-Stream to Popcount



#### • DMA

 Add DMA engine to move data via AXI4-Full to AXI-Stream interface

#### Popcount.sv:

- Add AXI-Stream Interface
- Keep AXI4-Lite Interface to read result

## Example DMA

https://pynq.readthedocs.io/en/v2.6.1/pynq\_libraries/dma.html

```
import numpy as np
from pynq import allocate
from pynq import Overlay
overlay = Overlay('example.bit')
dma = overlay.axi_dma
                                                                       Count (muto)
input_buffer = allocate(shape=(5,), dtype=np.uint32)
output buffer = allocate(shape-(5,), dtype-np.uint32)
for i in range(5):
   input buffer[i] = i
dma.sendchannel.transfer(input_buffer)
-dma-recvchannel_transfer(output_buffer)
dma.sendchannel.wait()
-dma.recvchannel.wait()
Output buffer will contain: [0 1 2 3 4]
```

### P6 – DMA from C

Start the MM2S channel running by setting the run/stop bit to 1 (MM2S\_DMACR.RS =

 The halted bit (DMASR.Halted) should deassert indicating the MM2S channel is running.

### 2. Skip

- 3. Write a valid source address to the MM2S\_SA register.
- 4. Write the number of bytes to transfer in the MM2S\_LENGTH register.

  The MM2S\_LENGTH register must be written last.
- 5. Poll MM2S\_DMASR.Idle bit for completion wat until MM25\_DMASR. Tale = (

# P7+ Accelerate Machine Learning

• Goal: Accelerate reference neural network

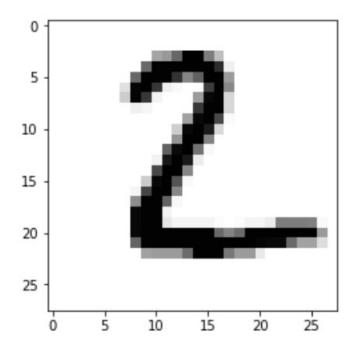
Harder, more open-ended projects

Groups of 2 allowed.

# Simple Neural Network

\_\_\_\_\_

Index: 0
Image:



ML Classification Result: 2
Real Value: 2

Correct Result: True

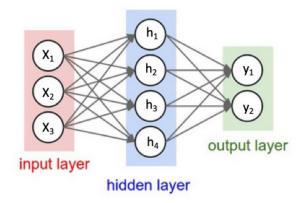
Correct Result: True

• Takes in image of number

Returns integer value

How? artificial neural network

# Why Dot Product?



```
# forward-pass of a 3-layer neural network:
f = lambda x: 1.0/(1.0 + np.exp(-x)) # activation function (use sigmoid)
x = np.random.randn(3, 1) # random input vector of three numbers (3x1)
h1 = f(np.dot(W1, x) + b1) # calculate first hidden layer activations (4x1)
h2 = f(np.dot(W2, h1) + b2) # calculate second hidden layer activations (4x1)
out = np.dot(W3, h2) + b3 # output neuron (1x1)
```

# Matrix Multiplication (Dot Product)

$$\begin{bmatrix} i_0 & i_1 \end{bmatrix} \times \begin{bmatrix} \omega_{00} & \omega_{10} & \omega_{20} \\ \omega_{01} & \omega_{11} & \omega_{21} \end{bmatrix} = \begin{bmatrix} 0_0 & 0_1 & 0_2 \end{bmatrix}$$

$$\begin{bmatrix} 0_0 & i_0 & \omega_{10} & \omega_{11} \\ 0_0 & i_0 & \omega_{11} & \omega_{21} \end{bmatrix}$$

$$\begin{bmatrix} 0_0 & i_0 & \omega_{10} & i_0 \\ 0_1 & i_0 & \omega_{10} & i_0 \end{bmatrix}$$

$$\begin{bmatrix} 0_1 & i_0 & \omega_{10} & i_0 \\ 0_2 & i_0 & \omega_{20} & i_0 \end{bmatrix}$$

$$\begin{bmatrix} 0_1 & i_0 & \omega_{10} & \omega_{21} \\ 0_0 & i_0 & \omega_{21} \end{bmatrix}$$

# Alternative Dot Computations

$$\begin{bmatrix}
 0.1 & 0.2 \end{bmatrix} \times \begin{bmatrix}
 1 & 2 & 3 \\
 4 & 5 & 6
\end{bmatrix} = \begin{bmatrix}
 0.9 & 1.2 & 1.5
\end{bmatrix}$$

$$\begin{bmatrix}
 0.1 & 0.2 & 0.1 \\
 0.2 & 0.3
\end{bmatrix}$$

$$\begin{bmatrix}
 0.2 & 0.2 & 0.2 \\
 0.2 & 0.2 & 0.3
\end{bmatrix}$$

# Alternative Dot Computations

$$\begin{bmatrix}
0.1 & 0.2
\end{bmatrix} \times \begin{bmatrix}
1 & 2 & 3\\
4 & 5 & 6
\end{bmatrix} = \begin{bmatrix}
0.9 & 1.2 & 1.5
\end{bmatrix}$$

$$\begin{bmatrix}
0 & 1.1 & 0.1.2 & 0.1.3
\end{bmatrix} + \begin{bmatrix}
0 & 0 & 0
\end{bmatrix} = \begin{bmatrix}
0.1 & 0.2 & 0.3
\end{bmatrix}$$

$$\begin{bmatrix}
0.1.1 & 0.2.5 & 0.2.6 + \begin{bmatrix}
0.1 & 0.2 & 0.3
\end{bmatrix} = \begin{bmatrix}
0.9 & (.2 & 1.5)
\end{bmatrix}$$

11

```
Input Weights Output
[[0.1 0.2 0.3]] . [1. 2. 3. 4.] = [3.8000002 4.4
[5. 6. 7. 8.]
[9. 10. 11. 12.]
```

### Alternative Dot

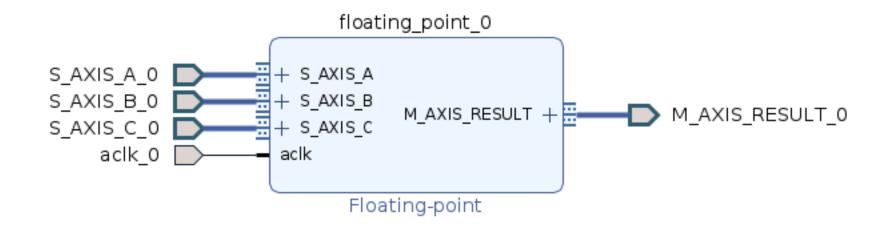
```
# how its done in dot.sv

def pydot(inputs, weights):
    inputs = inputs[0] # remove outer nesting
    outs = np.zeros(weights.shape[1], dtype=np.float32)
    for i in range(weights.shape[0]): # input length
        for j in range(weights.shape[1]): # output length
            outs[j] = outs[j] + weights[i][j] * inputs[i]
    return outs
```

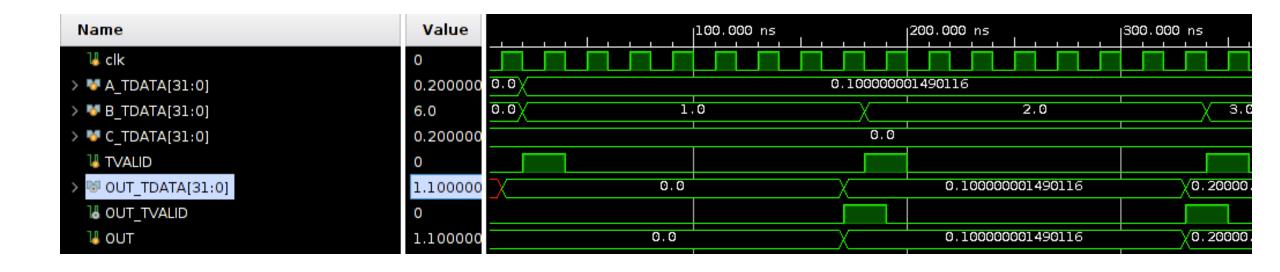
5. 5.6000004]

```
Inputs (Shape):
  (1, 3)
Output (Shape):
  (1, 4)
Weights (Shape):
  (3, 4)
```

### Hardware Dot Product



### Demo Time



# Floating-Point Multiply-Accumulate (FMAC)

• Math: a \* b + c 8 cycles

# Floating-Point math takes 8 cycles.

• Floating-Point is complicated.

How do we work around an 8 cycle latency?

Pipelining!

# Pipelining

- FMAC takes 8 cycles for 1 value
- But can accept a new value every cycle.

- Latency: 8 cycles / value
- Throughput: 1 value / cycle

# Latency vs. Throughput

 Latency: How long does an individual operation take?

• Throughput: How many operations can you complete in a given time?

# Pipelining

# Pipelined Dot Computations

$$[0.1 \ 0.2] \times [1 \ 2 \ 3] = [0.9 \ 1.2 \ (.5]$$

# Problems with Pipelined FMAC

Latency!

# Latency on Pipelined FMAC

• Solution: Stall at the end of a row.

• Drain the pipeline.

### Hardware Parallelism

```
• CPU: 1 Floating-Point Unit
• FPGA? (1) Floating-Point Units?
20?
100?
```

# Finding Parallelism

 Some some computation that doesn't depend on other computation's results

Shared Inputs are OK.

Next Time: Can we use 2+ FMACs?

reset por count in Hw?

-) reset w/ mmTo

$$\begin{bmatrix}
0.1 & 0.2 & 0.3
\end{bmatrix}
\begin{bmatrix}
1 & 2 & 3 & 4 \\
9 & 10 & 11 & 12
\end{bmatrix}
= \begin{bmatrix}
3.8 & 4.4 & 5 & 5.6
\end{bmatrix}$$

$$\begin{bmatrix}
0.1 & . & [ 2 & 3 & 4 ] + [ 0 & 0 & 0 & 0 ] = [ 0.1 & 0.2 & 0.3 & 0.4 ]
\end{bmatrix}$$

$$0.2 - [5 & 6 & 78] + [ 0.1 & 0.2 & 0.3 & 0.4 ] = [ 1.1 & 1.4 & 1.7 & 2.0 ]$$

$$0.3 - [9 & 10 & 11 & 1.4 & 1.7 & 2.0] = [ 3.8 & 4.4 & 5 & 5.6 ]$$

# Parallelize Alternative Dot Computations?

[0.1 0.2] × 
$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$$
 =  $\begin{bmatrix} 0.9 & 1.2 & (.5) \end{bmatrix}$   
[0.1 \cdot [1 & 23] +  $\begin{bmatrix} 0 & 0 & 0 \end{bmatrix}$  =  $\begin{bmatrix} 0 & 1 & 0.2 & 0.3 \end{bmatrix}$  =  $\begin{bmatrix} 0.1 & 0.2 & 0.3 \end{bmatrix}$  =  $\begin{bmatrix} 0.1 & 0.2 & 0.3 \end{bmatrix}$  =  $\begin{bmatrix} 0.2 & 0.4 & 0.2 & 0.3 \end{bmatrix}$  =  $\begin{bmatrix} 0.8 & [0 & 1.2] + [0.1 & 0.2 & 0.3] = \begin{bmatrix} 0.9 & (.2 & (.5) \end{bmatrix}$ 

# Can we parallelize Dot?

TOIL 0.2 0.3] [1 2 3 4] = [3.8 4.4 5 5.6]
9 WH 12

$$\begin{bmatrix}
 0.1 & 0.2 & 0.3 \\
 5 & 6.7 & 8 \\
 9 & 10 & 11 & 12
 \end{bmatrix} = \begin{bmatrix}
 3.8 & 4.4 & 5 & 5.6 \\
 9 & 10 & 11 & 12
 \end{bmatrix}$$

$$\begin{bmatrix}
6.1 & 0.3 \\
9.101112
\end{bmatrix} = \begin{bmatrix}
2.8 & 3.2 & 3.6 & 4
\end{bmatrix}$$

$$\begin{bmatrix}
6.1 & 0.3 \\
9.101112
\end{bmatrix} = \begin{bmatrix}
1.2 & 1.4 & 1.6
\end{bmatrix}$$

# Can we parallelize Dot?

```
# how its done in dot.sv
def pydot(inputs, weights):
    inputs = inputs[0] # remove outer nesting
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    for i in range(weights.shape[0]): # input length
        for j in range(weights.shape[1]): # output length
            outs[j] = outs[j] + weights[i][j] * inputs[i]
    return outs
```

# Can we parallelize Dot?

# 19: Hardware Acceleration III

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