## Matrix Multiplication

## 1. HLS C-sim/Synthesis/Cosim (Screenshot + brief intro) (1%)

矩陣乘法是一種將兩個矩陣合併為第三個矩陣的二進制運算。運算本身可以描述為對兩個矩陣的向量作線性運算。矩陣乘法的最常見形式稱為矩陣乘積。當矩陣A的尺寸為nxm而矩陣B的尺寸為mxp時,矩陣乘積AB創建一個nxp矩陣。

$$\mathbf{A} = \begin{bmatrix} A_{11} & A_{12} & \cdots & A_{1m} \\ A_{21} & A_{22} & \cdots & A_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1} & A_{n2} & \cdots & A_{nm} \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} B_{11} & B_{12} & \cdots & B_{1p} \\ B_{21} & B_{22} & \cdots & B_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ B_{m1} & B_{m2} & \cdots & B_{mp} \end{bmatrix}$$

$$\mathbf{AB} = \begin{bmatrix} (\mathbf{AB})_{11} & (\mathbf{AB})_{12} & \cdots & (\mathbf{AB})_{1p} \\ (\mathbf{AB})_{21} & (\mathbf{AB})_{22} & \cdots & (\mathbf{AB})_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ (\mathbf{AB})_{n1} & (\mathbf{AB})_{n2} & \cdots & (\mathbf{AB})_{nn} \end{bmatrix}$$

where the operation  $(\mathbf{AB})_{ij}$  is defined as  $(\mathbf{AB})_{ij} = \sum_{k=1}^{m} A_{ik} B_{kj}$ .

### ◎圖 1 Matrix Multiplication Operation

```
1 #include "matrixmultiplication.h"
 3 void matrixmul(int A[N][M], int B[M][P], int AB[N][P]) {
   #pragma HLS ARRAY_RESHAPE variable=A complete dim=2
     #pragma HLS ARRAY_RESHAPE variable=B complete dim=1
     /* for each row and column of AB */
 6
     row: for(int i = 0; i < N; ++i) {
 7
        col: for(int j = 0; j < P; ++j) {</pre>
 9
          #pragma HLS PIPELINE II=1
          /* compute (AB)i,j */
10
          int ABij = 0;
11
        product: for(int k = 0; k < M; ++k) {
12
            ABij += A[i][k] * B[k][j];
13
         AB[i][j] = ABij;
15
16
17
18 }
```

◎圖 2 Baseline Code

The Screenshot is for the Baseline Code

#### Comparing against output data \*\*\*\*\*\*\*\*\*\*\*\* PASS: The output matches the golden output! \*\*\*\*\*\*\*\*\*\*\* INFO: [SIM 1] CSim done with 0 errors. ◎圖 3 C-sim **Performance Estimates** □ Timing **■** Summary Clock Target Estimated Uncertainty ap\_clk 10.00 ns 8.742 ns ■ Latency □ Summary Latency (cycles) Latency (absolute) Interval (cycles) min max min max min max Type 1030 10.300 us 10.300 us 1030 1030 none 1030 Detail **■** Instance + Loop **Utilization Estimates** □ Summary BRAM\_18K DSP48E LUT URAM Name DSP Expression 96 Ω 1804 FIFO Instance Memory Multiplexer 3757 Register 96 0

◎圖 4 C-Synthesis(需要非常多的 DSP48E → Bad)

43

3757

220 106400 53200

## **Cosimulation Report for 'matrixmul'**

280

Total

Available

Utilization (%)

Result							
		Latency			Interval		
RTL	Status	min	avg	max	min	avg	max
VHDL	NA	NA	NA	NA	NA	NA	NA
Verilog	Pass	1030	1030	1030	NA	NA	NA

Export the report(.html) using the Export Wizard

◎圖 5 Cosim

## 2. Improvement – throughput, area (1%)

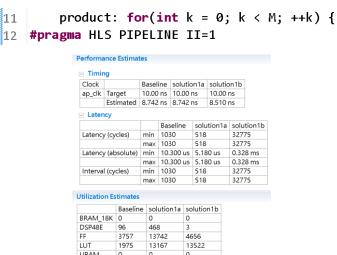
因為是矩陣乘法必然會有很多的乘法跟並行的運算,所以可以先從減少乘法(DSP48 →面積),以及利用 Pipeline 的運算來減少 Latency。

### Solution1: Change the location of the pipeline directive

#### 1a. Insert the pragma in row loop

```
7    row: for(int i = 0; i < N; ++i) {
8          #pragma HLS PIPELINE II=1</pre>
```

### 1b. Insert the pragma in inter loop



◎圖 6 Compare Report (Result shows the location of pipeline directive in Baseline is best).

In solution1a the use of DSP48E is more than can use.

## Solution2: Change the array reshape directives

#### 2a. Remove all array reshape directives

Unable to schedule 'load' operation on array 'A' due to limited memory ports. Please consider using a memory core with more ports or partitioning the array 'A'.

Unable to schedule 'load' operation on array 'B' due to limited memory ports. Please consider using a memory core with more ports or partitioning the array 'B'.

```
12     product: for(int k = 0; k < M; ++k) {
13           ABij += A[i][k] * B[k][j];
14      }
15      AB[i][j] = ABij;</pre>
```

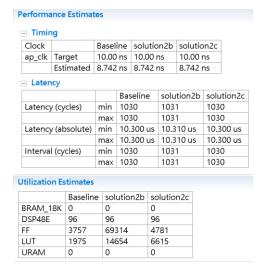
◎圖 7 因為 line14, 所以必須將 Array Reshape

#### 2b. Edit the dim

```
#pragma HLS ARRAY_RESHAPE variable=A complete dim=2
#pragma HLS ARRAY_RESHAPE variable=B complete dim=0
```

### 2c. Edit the dim

```
#pragma HLS ARRAY_RESHAPE variable=A complete dim=2
#pragma HLS ARRAY_PARTITION variable=B complete dim=0
```



◎圖 8 Compare Report(The improvement is fail)

#### **Solution3: Block Matrix Multiplication**

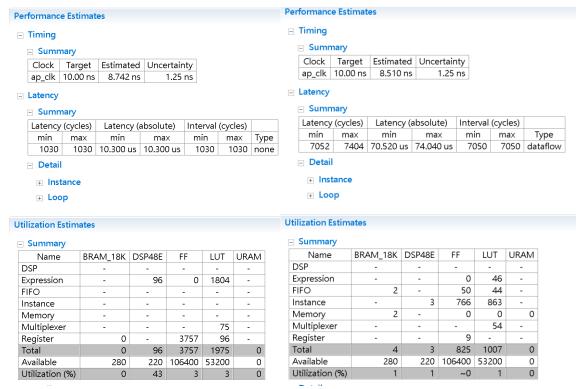
Separate origin matrix into submatrix:

$$\mathbf{A} = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$$
$$A_{11} = \begin{bmatrix} a & b \\ e & f \end{bmatrix}, A_{12} = \begin{bmatrix} c & d \\ g & h \end{bmatrix}, A_{21} = \begin{bmatrix} i & j \\ m & n \end{bmatrix}, A_{22} = \begin{bmatrix} k & l \\ o & p \end{bmatrix}$$

◎圖 9 Block based Matrix

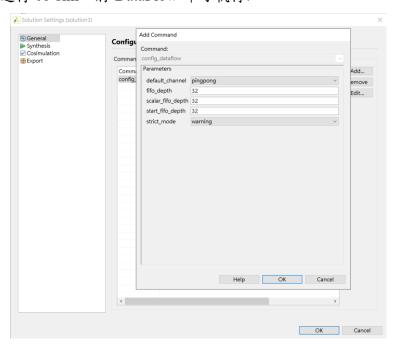
```
1 #include "block_mm.h"
 2@void blockmatmul(hls::stream<blockvec> &Arows, hls::stream<blockvec> &Bcols,
             blockmat &ABpartial, int it) {
 4 #pragma HLS DATAFLOW
      int counter = it % (SIZE/BLOCK_SIZE);
       static DTYPE A[BLOCK_SIZE][SIZE];
       if(counter == 0){ //only load the A rows when necessary
         loadA: for(int i = 0; i < SIZE; i++) {</pre>
 9
           blockvec tempA = Arows.read();
 10
11
           for(int j = 0; j < BLOCK_SIZE; j++) {</pre>
              #pragma HLS PIPELINE II=1
12
              A[j][i] = tempA.a[j];
13
14
15
       DTYPE AB[BLOCK_SIZE][BLOCK_SIZE] = { 0 };
17
18
       partialsum: for(int k=0; k < SIZE; k++) {</pre>
19
         blockvec tempB = Bcols.read();
         for(int i = 0; i < BLOCK_SIZE; i++) {
  for(int j = 0; j < BLOCK_SIZE; j++) {
    AB[i][j] = AB[i][j] + A[i][k] * tempB.a[j];</pre>
20
21
22
23
24
        }
25
       writeoutput: for(int i = 0; i < BLOCK_SIZE; i++) {
26
27
         for(int j = 0; j < BLOCK_SIZE; j++) {</pre>
           ABpartial.out[i][j] = AB[i][j];
28
29
30
31 }
```

◎圖 10 block mm.cpp



◎圖 11 Compare Report(Now just need 1DSP48E)

在原本的 code 會無法進行 co-sim,將 DataFlow 即可執行:



◎圖 12 DataFlow 問題修正

# 3. Github submit (1%)

https://github.com/r08943099/MSOCFall2020

4. Complexity (1% - Instructor/ TA decide)