

# Matrix Multiplication

Speaker: Jia-Ming Lin

# Outline

- Matrix Multiplication and Applications
- Complete Matrix Multiplication
  - Smaller size, e.g.  $32 \times 32$
- Block Matrix Multiplication
  - Larger size, e.g.  $1024 \times 1024$

# Matrix Multiplication and Applications

# Matrix Multiplication

$$\begin{matrix} & A & & B & & C \\ \begin{matrix} 2 & 3 & 1 & 4 \\ 5 & 3 & 1 & 2 \\ 3 & 2 & 4 & 5 \\ 1 & 1 & 2 & 6 \end{matrix} & \times & \begin{matrix} 1 & 2 & 1 & 2 \\ 1 & 2 & 3 & 2 \\ 2 & 1 & 4 & 6 \\ 1 & 4 & 5 & 4 \end{matrix} & = & \begin{matrix} & & & \\ & & 28 & \\ & & & \\ & & & \end{matrix} \end{matrix}$$

$C[i][j] = \text{inner product of } A[i][\dots] \text{ and } B[\dots][j]$

# Application

- Computation of Neural Network
  - Fully connected layer

2	3	1	4
5	3	1	2
3	2	4	5
1	1	2	6

**Model Parameters**

×

1
3
4
5

**Feature**

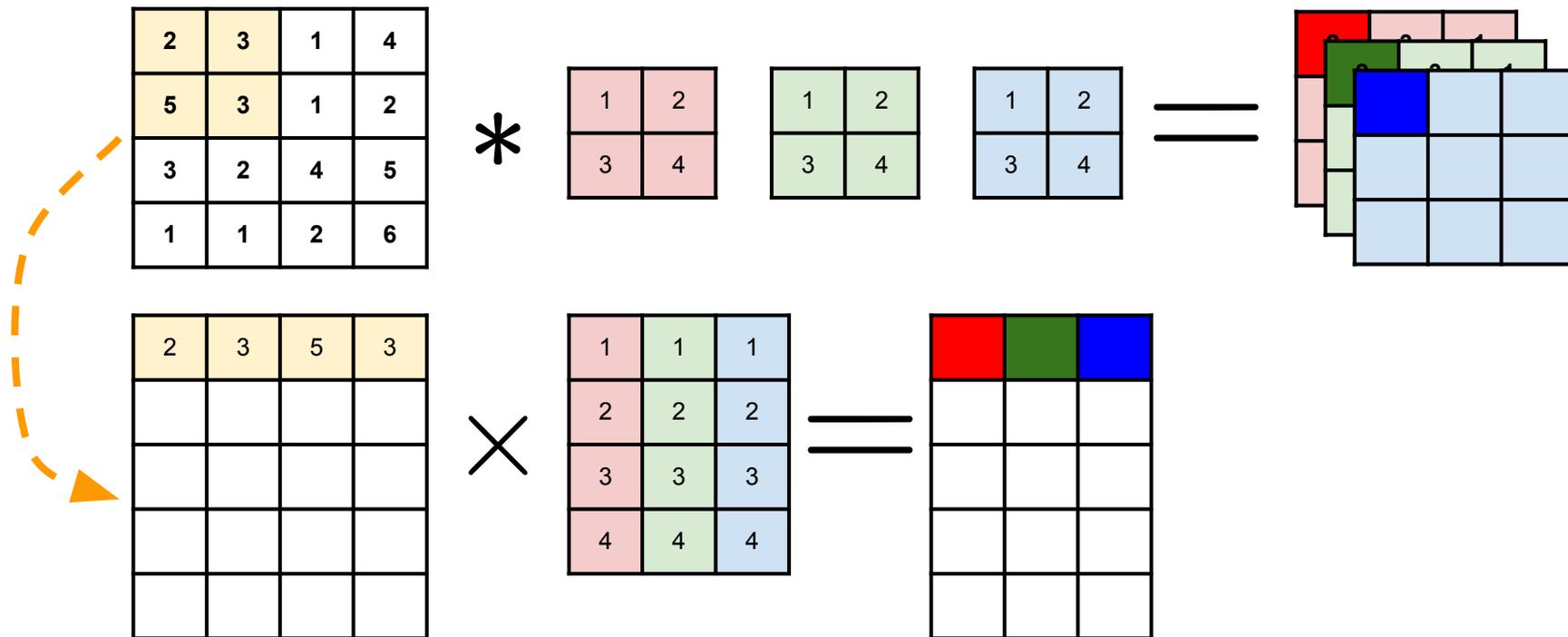
=

28

**Output**

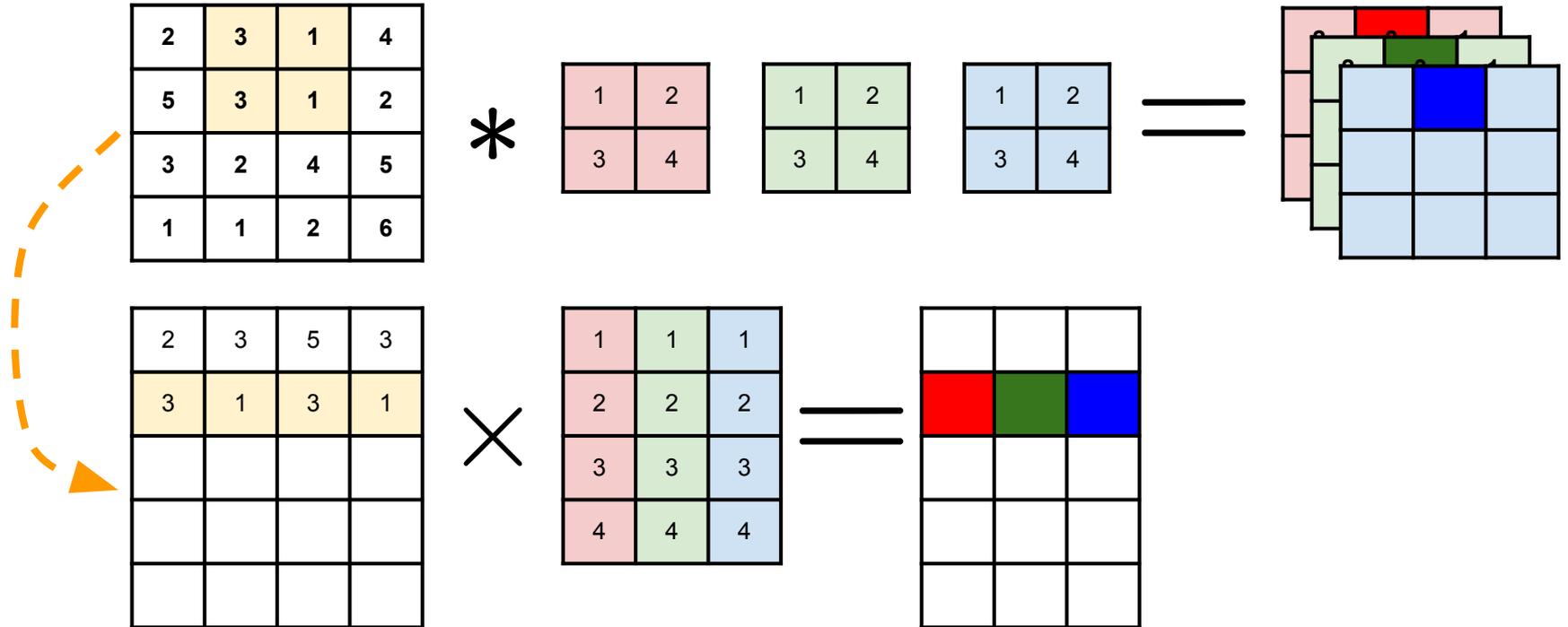
# Application

- Computation of Neural Network
  - Convolution Layer



# Application

- Computation of Neural Network
  - Convolution Layer



# Memory Hierarchy

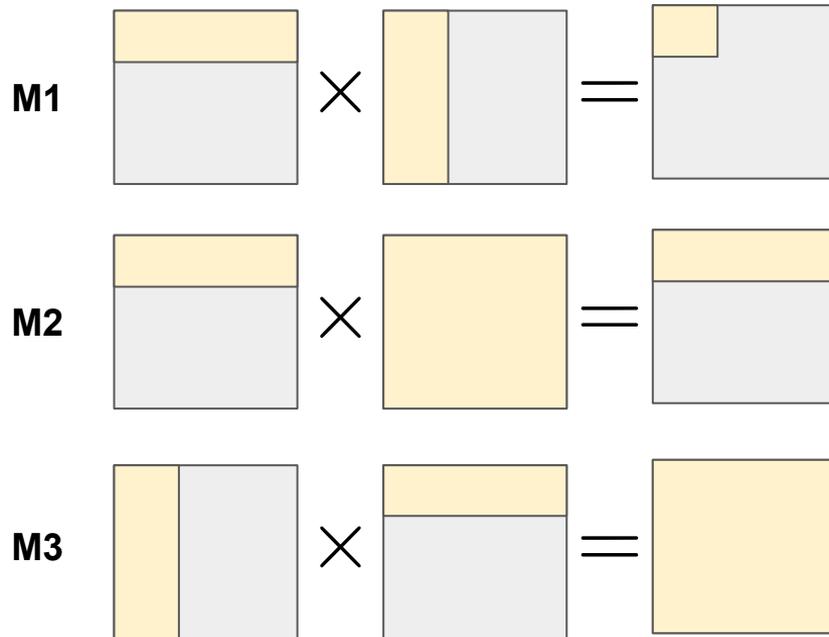
	External Memory	BRAM	FFs
count	1-4	thousands	millions
size	GBytes	KBytes	Bits
total size	GBytes	MBytes	100s of KBytes
width	8-64	1-16	1
total bandwidth	GBytes/sec	TBytes/sec	100s of TBytes/sec

- **External Memory:** highest density, lowest bandwidth
- **FFs:** highest total bandwidth, limited amount of data storage capability.
- **BRAM:** intermediate value between external memory and FFs

# Complete Matrix Multiplication

# Methodologies

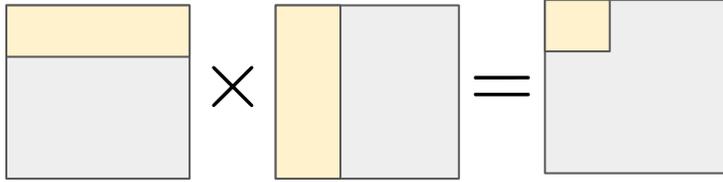
- Comparison
  - Latency, Resource Consumption



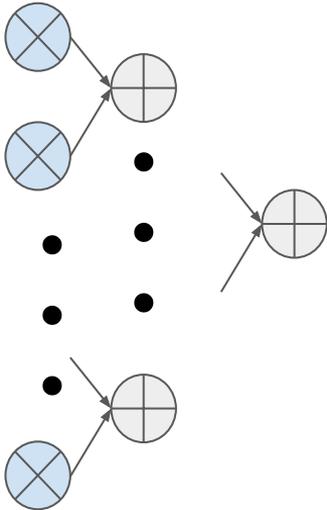
## Suppose

- Matrix size = 32x32
- Numbers in 16-bit integer
- Partial sum in 32-bit integer

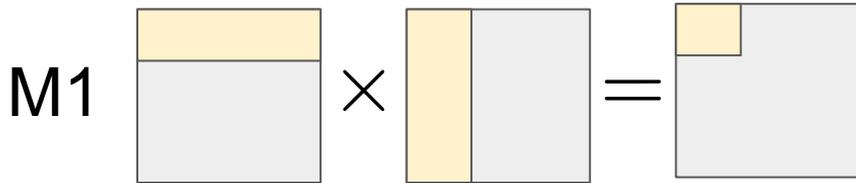
# M1



- Compute one inner product in **adder tree**
  - One output: need  $\log(32) = 5$  cycles



```
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 */  
    row: for(int i = 0; i < N; ++i) {  
        col: for(int j = 0; j < P; ++j) {  
            #pragma HLS PIPELINE II=1  
            /* compute (AB)ij */  
            int ABij = 0;  
            product: for(int k = 0; k < M; ++k) {  
                ABij += A[i][k] * B[k][j];  
            }  
            AB[i][j] = ABij;  
        }  
    }  
}
```



- Latency:

- Pipeline Depth = 5
- # of Iterations = 32x32
- Initialization Interval = 1
- Latency = 1024+4 = 1028

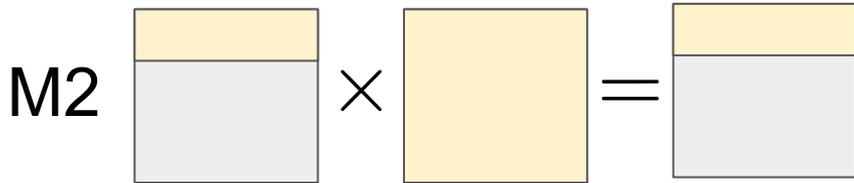
- Resource Consumption

- DSP = 32

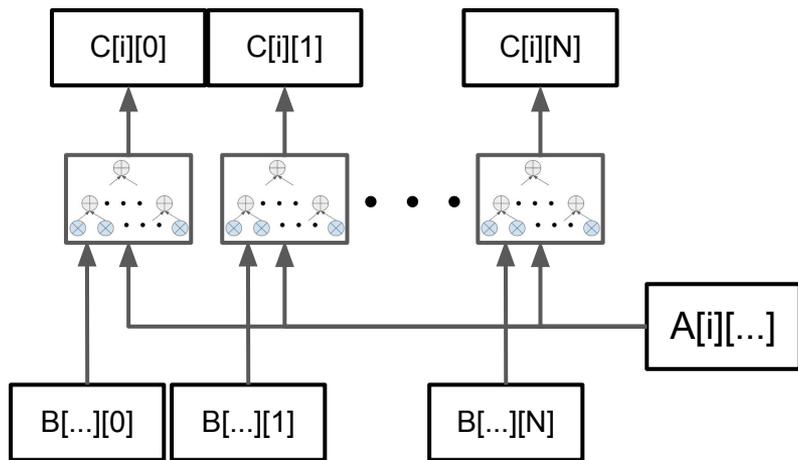
```

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 */
    row: for(int i = 0; i < N; ++i) {
        col: for(int j = 0; j < P; ++j) {
            #pragma HLS PIPELINE II=1
            /* compute (AB)i,j */
            int ABij = 0;
            product: for(int k = 0; k < M; ++k) {
                ABij += A[i][k] * B[k][j];
            }
            AB[i][j] = ABij;
        }
    }
}

```

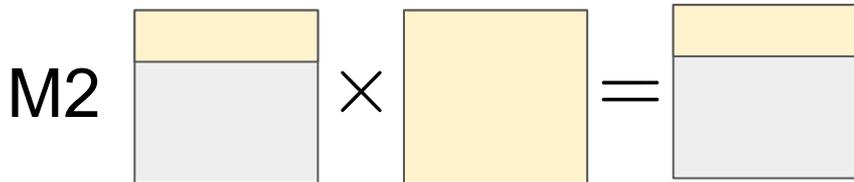


- Multiple **parallel adder trees**
  - One output: need  $\log(32) = 5$  cycles



```
void inner_product(A_i, B_j, C_ij){
#pragma HLS INLINE off
    for(k = 0...N){ // Unroll
        C_ij += A_i[k] * B_j[k]
    }
}
```

```
void mat_mul(A[N][N], B[N][N], C[N][N]){
#pragma HLS ALLOCATION instances=inner_product
limit=32 function
    for(i = 0...N){
        for(j = 0...N){ // Unroll
            inner_product(A[i][...], B[...][j], C[i][j])
        }
    }
}
```



- Latency:

- Pipeline Depth = 5
- # of Iterations = 32
- Initialization Interval = 5
- Latency =  $32 * 5 = 160$  (cycles)

- Resource Consumption

- DSP =  $32 * 32 = 1024$

```

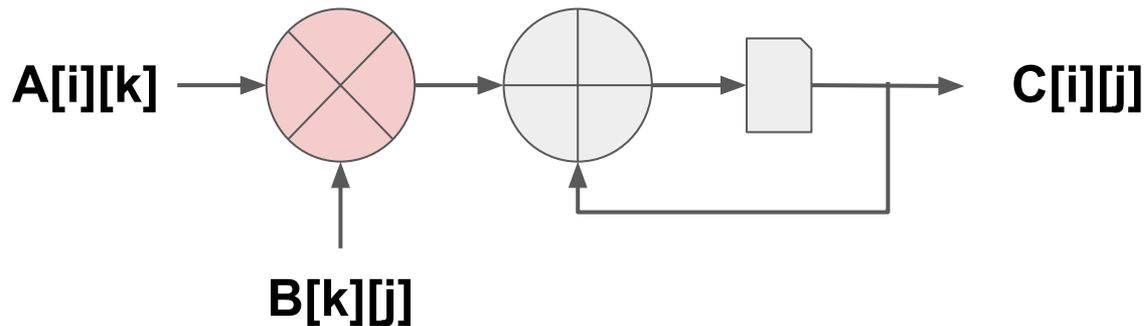
void inner_product(A_i, B_j, C_ij){
#pragma HLS INLINE off
    for(k = 0...N){ // Unroll
        C_ij += A_i[k] * B_j[k]
    }
}

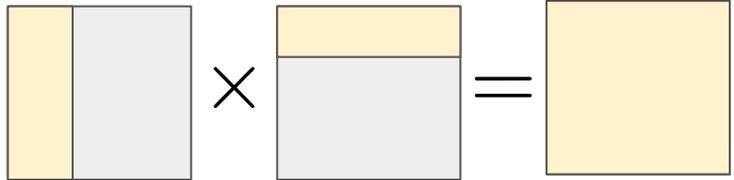
void mat_mul(A[N][N], B[N][N], C[N][N]){
#pragma HLS ALLOCATION instances=inner_product
limit=32 function
    for(i = 0...N){
        for(j = 0...N){ // Unroll
            inner_product(A[i][...], B[...][j], C[i][j])
        }
    }
}

```

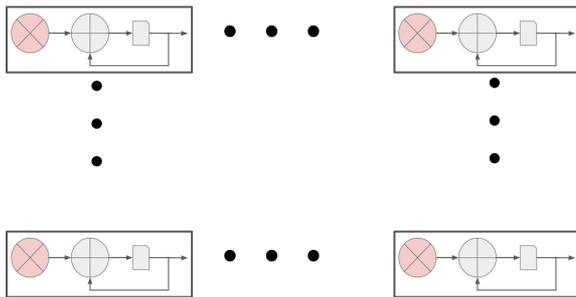


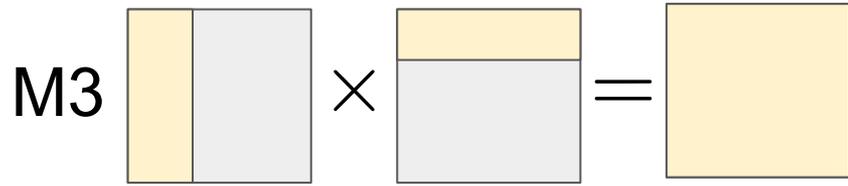
- Parallelize Partial Sum Computations
  - Parallelized outer product



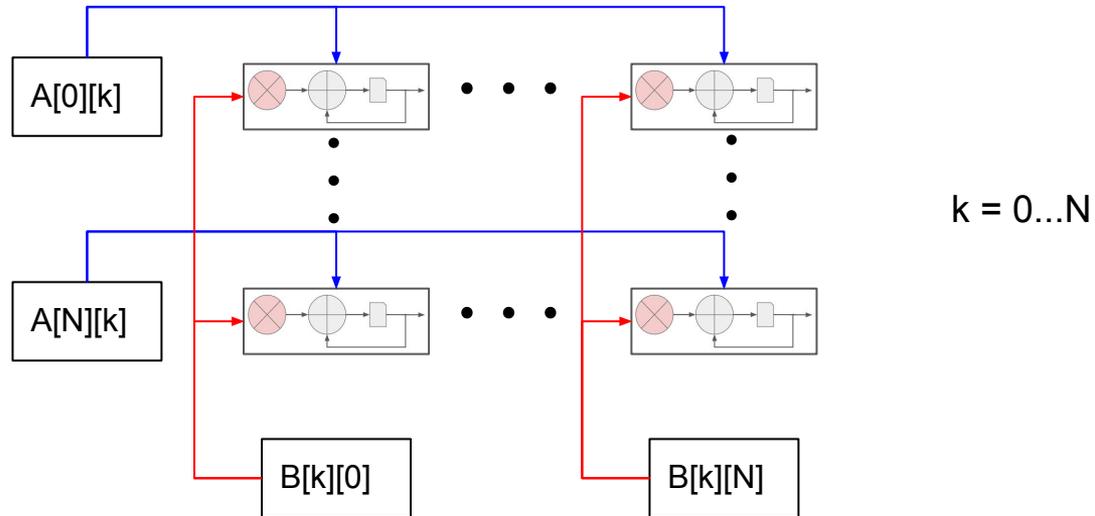
M3 

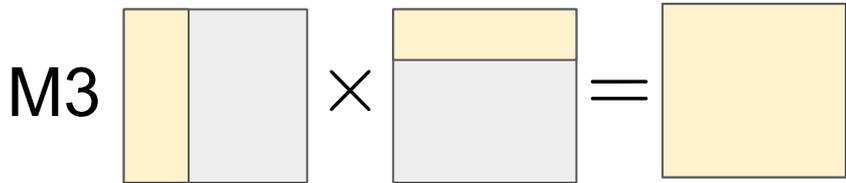
- Parallelize Partial Sum Computations



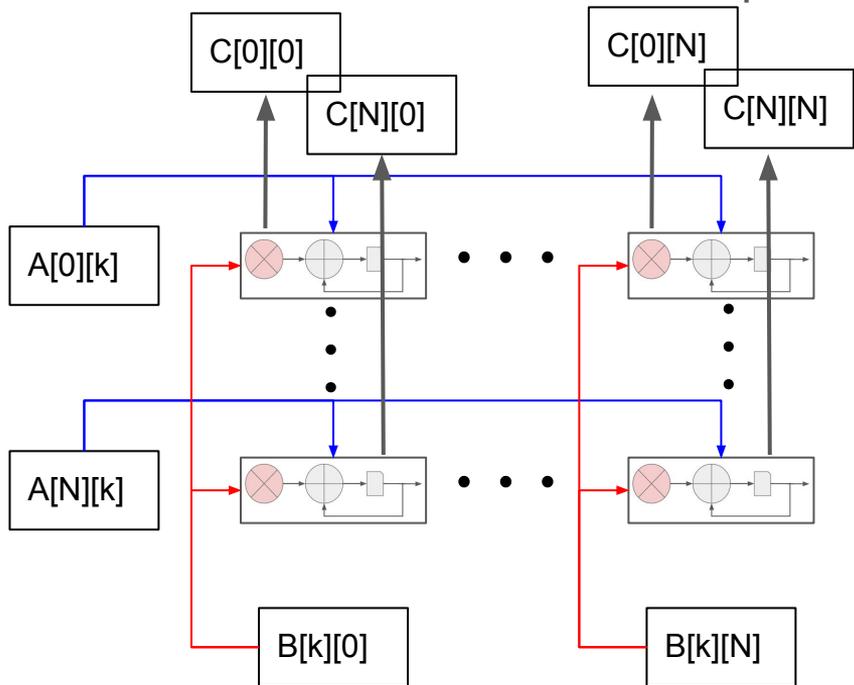


- Parallelize Partial Sum Computations





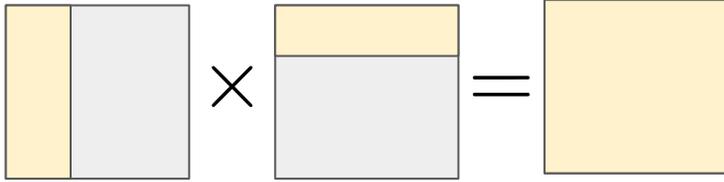
- Parallelize Partial Sum Computations



```
int32 PE(a, b, c){
    return a*b+c
}
```

```
void mat_mul(A[N][N], B[N][N], C[N][N]){
    #pragma HLS ALLOCATION instances=PE limit=1024 function
    for(k=0...N){ // pipeline
        A_col = A[...][k];
        B_row = B[k][...];
        for(i = 0...N){ // unroll
            for(j = 0...N){ // unroll
                C[i][j] = PE(A_col[i], B_row[j], C[i][j]);
            }
        }
    }
}
```

# M3



- Latency:
  - Pipeline Depth = 2
  - # of iterations = 32
  - Initialize interval = 1
  - Latency = 1+32 = 33 (cycles)
- Resource Consumption
  - DSP = 32\*32 = 1024

```
int32 PE(a, b, c){
    return a*b+c
}

void mat_mul(A[N][N], B[N][N], C[N][N]){
#pragma HLS ALLOCATION instances=PE limit=1024 function
    for(k=0...N){ // pipeline
        A_col = A[...][k];
        B_row = B[k][...];
        for(i = 0...N){ // unroll
            for(j = 0...N){ // unroll
                C[i][j] = PE(A_col[i], B_row[j], C[i][j]);
            }
        }
    }
}
```