

# PARALLEL MATRIX MULTIPLICATION

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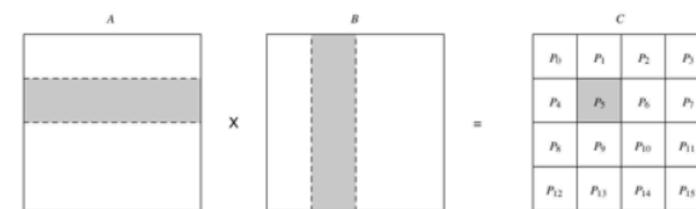
# Outline

- Problem Statement
- Applications
- Sequential Matrix Multiplication
- Parallel Matrix Multiplication
- Cannon's Algorithm
- Numerical example
- Results
- Next Steps



# PROBLEM STATEMENT

If matrix  $A = [a_{ij}]$  is an  $m \times n$  matrix and  $B = [b_{ij}]$  is an  $n \times p$  matrix then the matrix multiplication  $A \times B$  is an  $m \times p$  matrix.

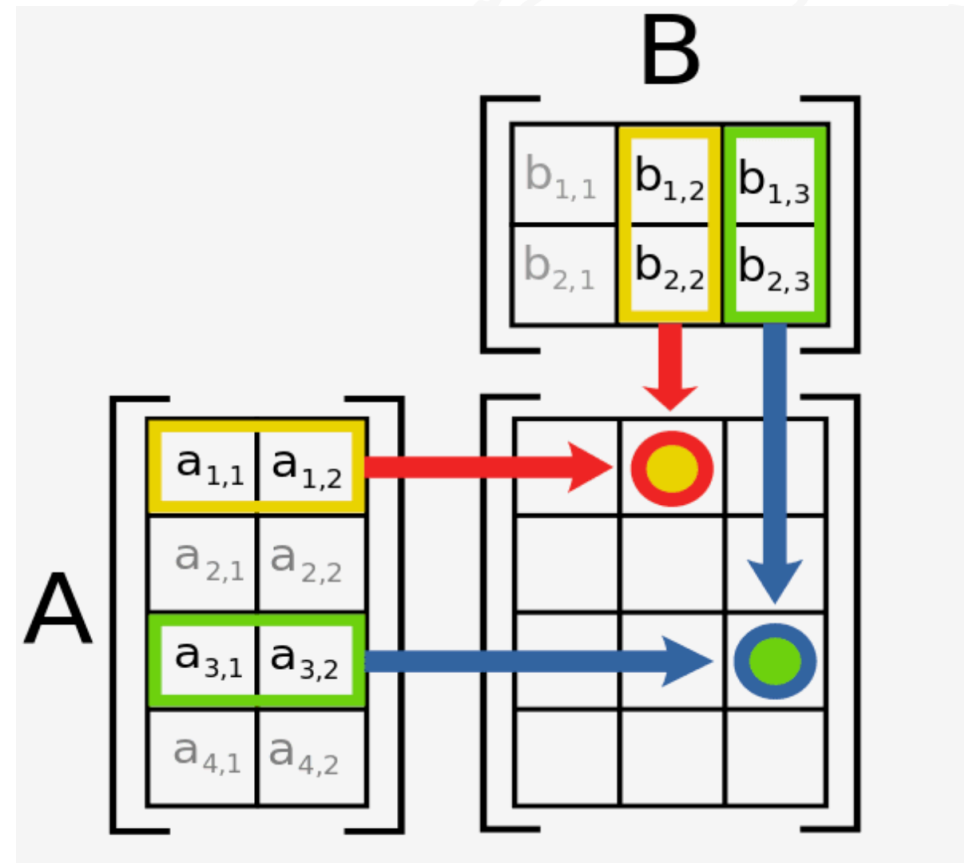


$$AB = [c_{ij}], \text{ where } c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{in}b_{nj}$$

(The entry in the  $i$ th row and  $j$ th column is denoted by the double subscript notation  $a_{ij}$ ,  $b_{ij}$ , and  $c_{ij}$ .)

## PROBLEM STATEMENT (contd.)

Number of Columns in A match the number of Rows of B.



Depiction of matrix multiplication, taken from [Wikipedia](#)

# Applications

- Used in image filtering using 2D convolution
- Used in Machine learning algorithms
- Used for quantum mechanics

Image Patch 1	Image Patch 2	Image Patch 3
[ 1., 2., 3., 4.]	[ 1., 2., 3., 4.]	[ 1., 2., 3., 4.]
[ 5., 6., 7., 8.]	[ 5., 6., 7., 8.]	[ 5., 6., 7., 8.]
[ 9., 10., 11., 12.]	[ 9., 10., 11., 12.]	[ 9., 10., 11., 12.]
[13., 14., 15., 16.]	[13., 14., 15., 16.]	[13., 14., 15., 16.]
Image Patch 4	Image Patch 5	Image Patch 6
[ 1., 2., 3., 4.]	[ 1., 2., 3., 4.]	[ 1., 2., 3., 4.]
[ 5., 6., 7., 8.]	[ 5., 6., 7., 8.]	[ 5., 6., 7., 8.]
[ 9., 10., 11., 12.]	[ 9., 10., 11., 12.]	[ 9., 10., 11., 12.]
[13., 14., 15., 16.]	[13., 14., 15., 16.]	[13., 14., 15., 16.]
Image Patch 7	Image Patch 8	Image Patch 9
[ 1., 2., 3., 4.]	[ 1., 2., 3., 4.]	[ 1., 2., 3., 4.]
[ 5., 6., 7., 8.]	[ 5., 6., 7., 8.]	[ 5., 6., 7., 8.]
[ 9., 10., 11., 12.]	[ 9., 10., 11., 12.]	[ 9., 10., 11., 12.]
[13., 14., 15., 16.]	[13., 14., 15., 16.]	[13., 14., 15., 16.]

All the possible  $2 \times 2$  image patches in X given the parameters of the 2D convolution. Each color represents a unique patch

[ 1., 2., 3., 5., 6., 7., 9., 10., 11.]
[ 2., 3., 4., 6., 7., 8., 10., 11., 12.]
[ 5., 6., 7., 9., 10., 11., 13., 14., 15.]
[ 6., 7., 8., 10., 11., 12., 14., 15., 16.]

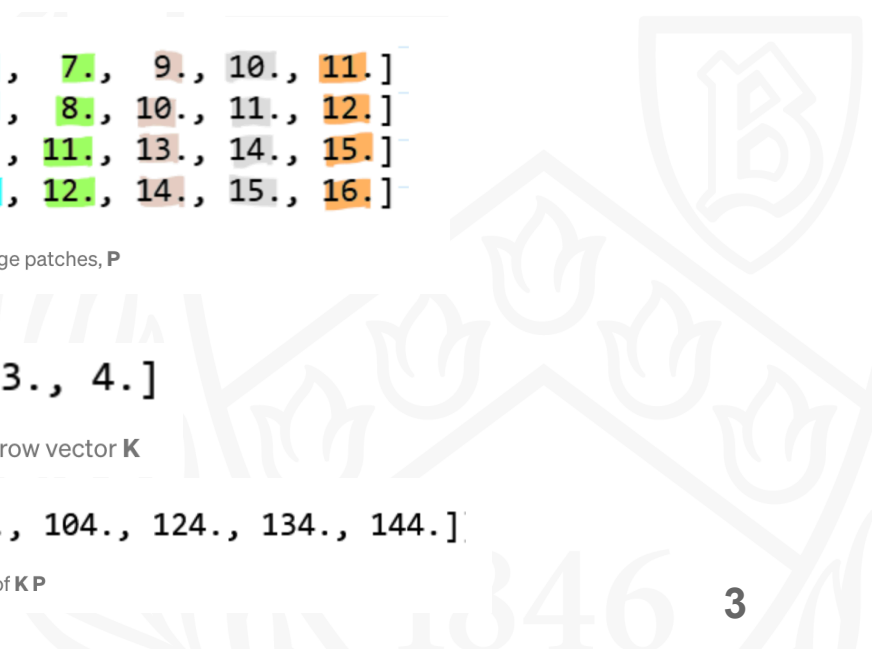
The matrix of image patches, P

[1., 2., 3., 4.]

W as a flattened row vector K

[ 44., 54., 64., 84., 94., 104., 124., 134., 144.]

The result of KP



# Sequential Matrix Multiplication

```
for (i=0; i< n; i++){  
    for (j=0; j< n; j++){  
        c[i][j] = 0  
        for (k=0; k< n; k++){  
            c[i][j] = c[i][j] + a[i][k]*b[k][j]  
        }  
    }  
}
```

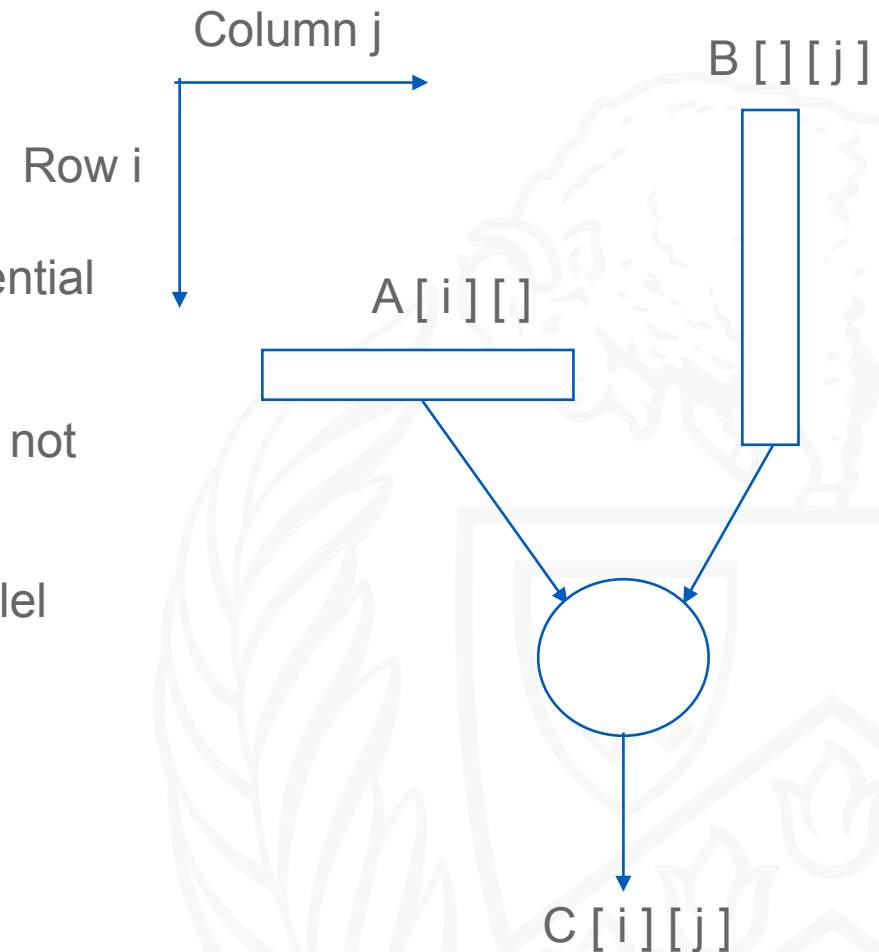


**NOTE:** The total number of steps are  $n*n*(2n-1)$ . So the complexity of the problem is  $O(n^3)$



# Parallel Matrix Multiplication

- Parallel matrix multiplication is usually based on the sequential matrix multiplication algorithm.
- The computation in each iteration of the two outer loops is not dependent upon any other iteration.
- Each instance of the inner loop could be executed in parallel
- Complexity of  $O(n^2)$  is obtainable with  $n$  processors
- Complexity of  $O(n)$  is obtainable with  $n^2$  processors
- Complexity of  $O(\log n)$  is obtainable with  $n^3$  processors



# Parallel - Cannon's Algorithm

Step 1: Divide the matrix A and B into P square blocks, where P is the number of Processors.

Step 2: Create grid of processors of size  $P^{1/2} * P^{1/2}$  So that each process had block of A and block of B

Step 3: Each process has a sub block C to which we add the results after Multiplying the sub-blocks in the processor.

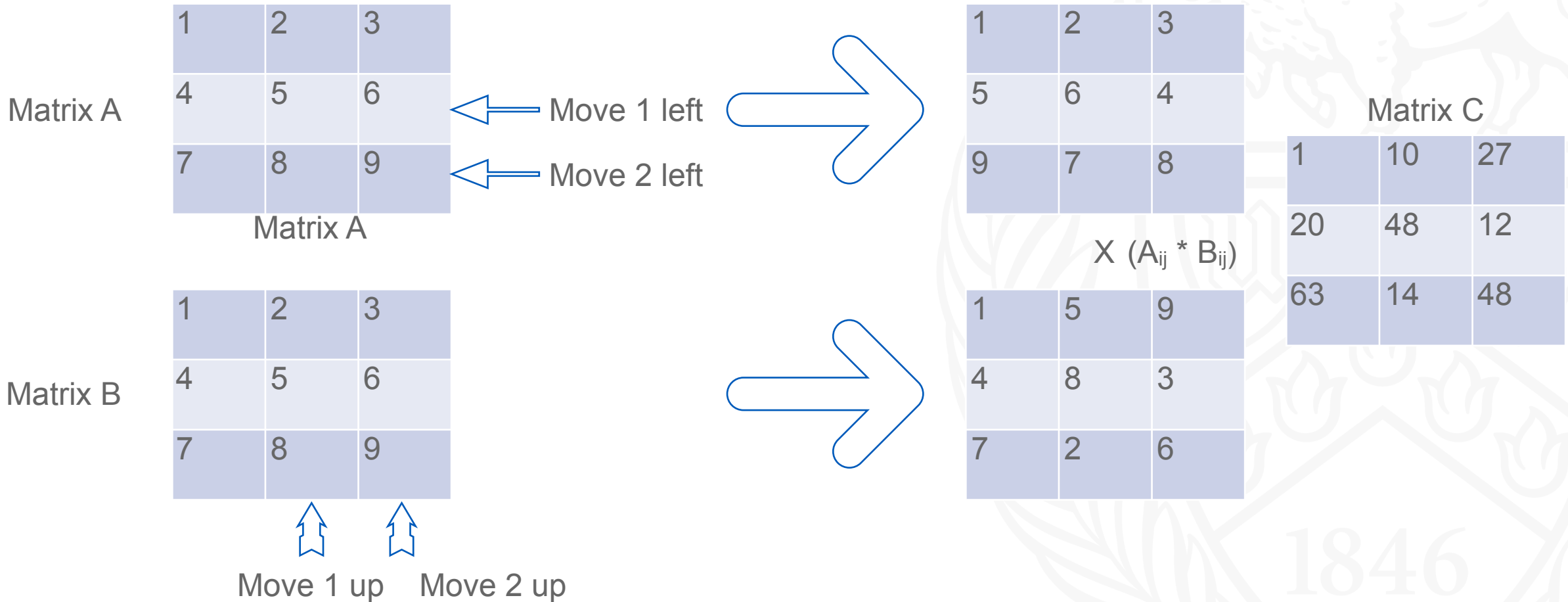
Step 4: The sub-blocks of A are shifted one step to the left and the sub-blocks of B are shifted one step up.

Step 5: We repeat the steps 3 and 4 for square root of P times.

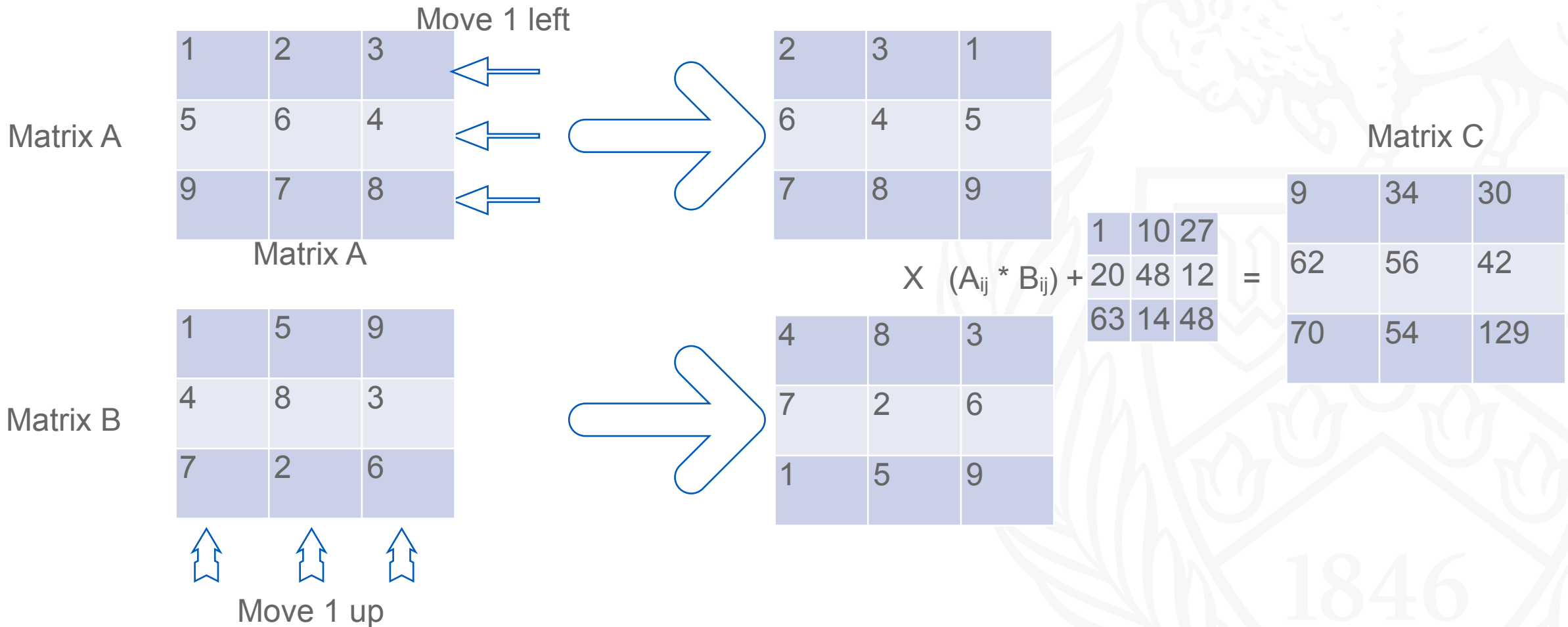




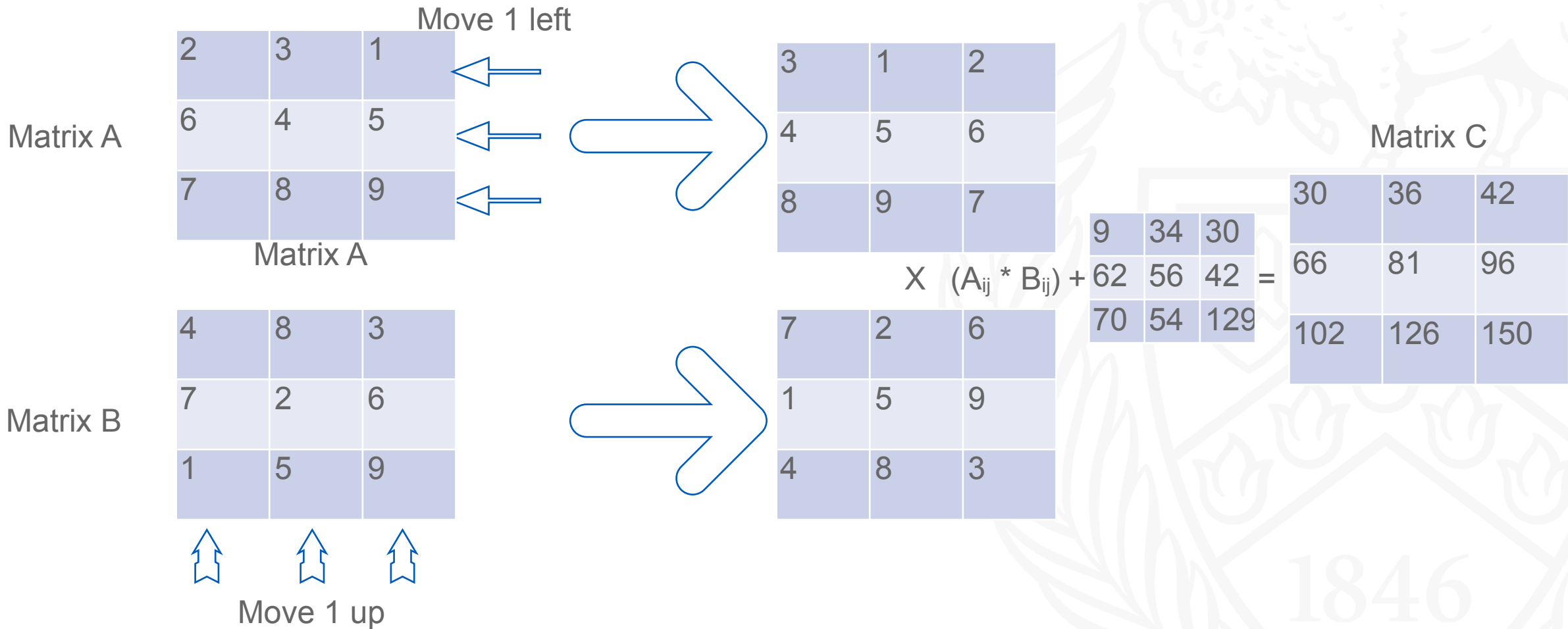
# Numerical Example



# Numerical Example (Contd.)

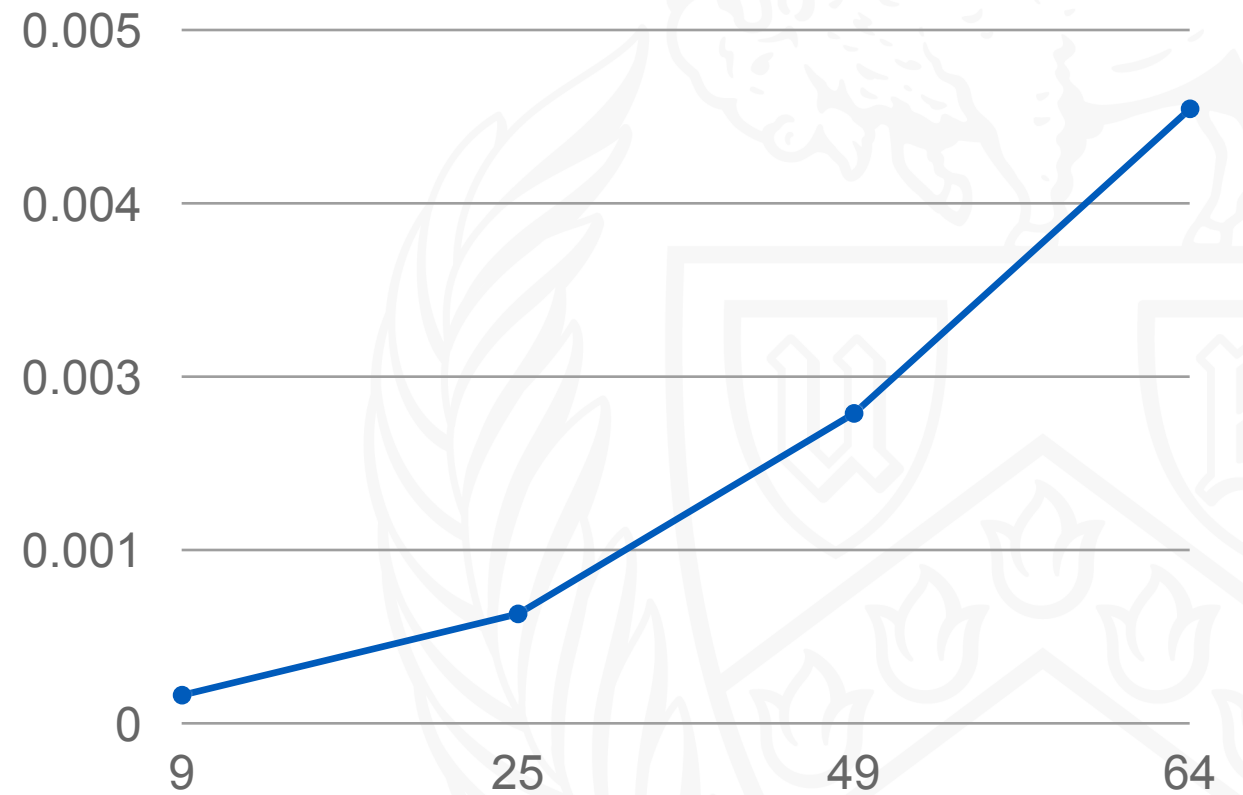


# Numerical Example (Contd.)



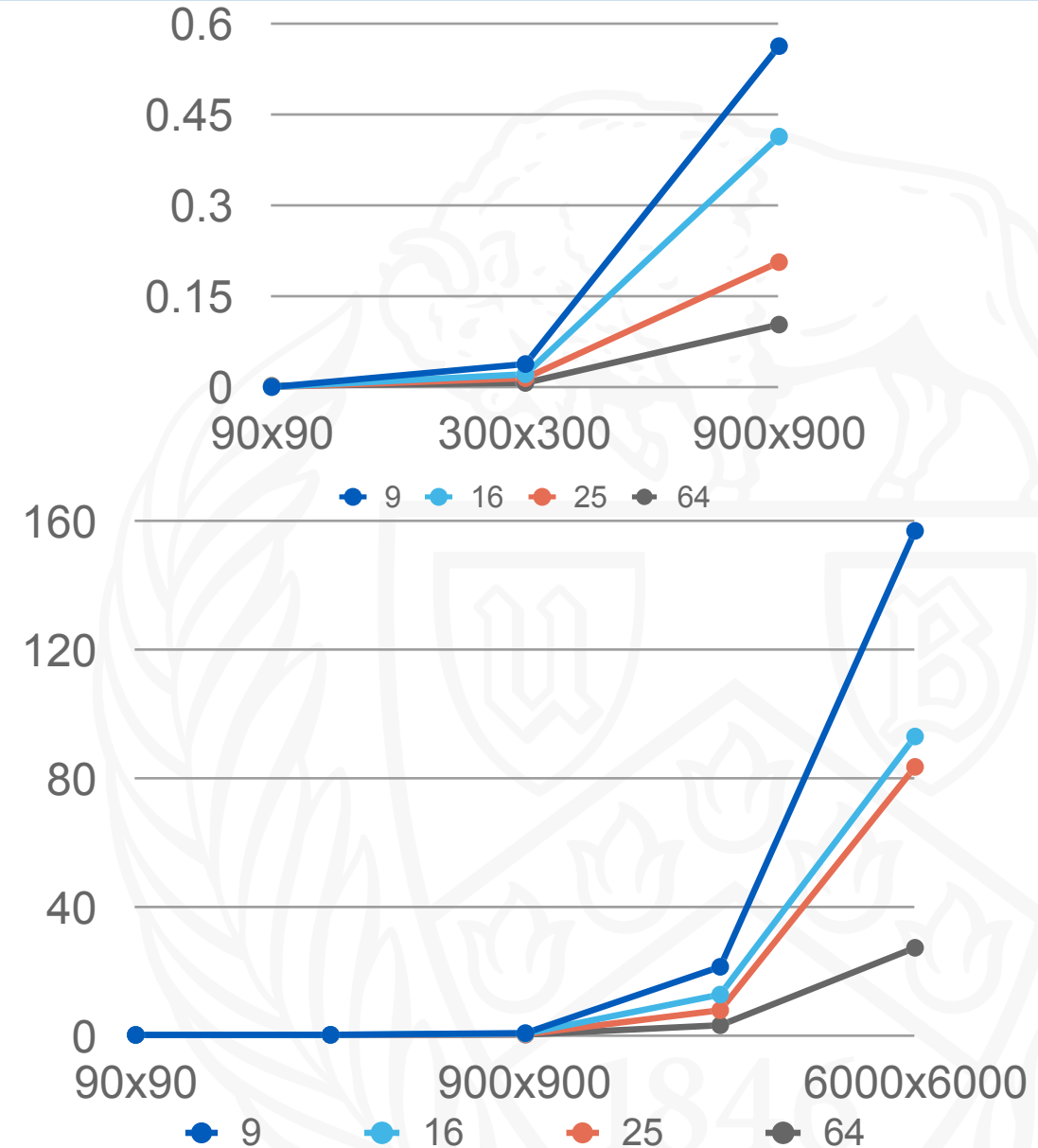
# Results

Matrix Size	No of processors	Time (secs)
3*3	9	0.000204
5*5	25	0.000791
7*7	49	0.002239
8*8	64	0.004439



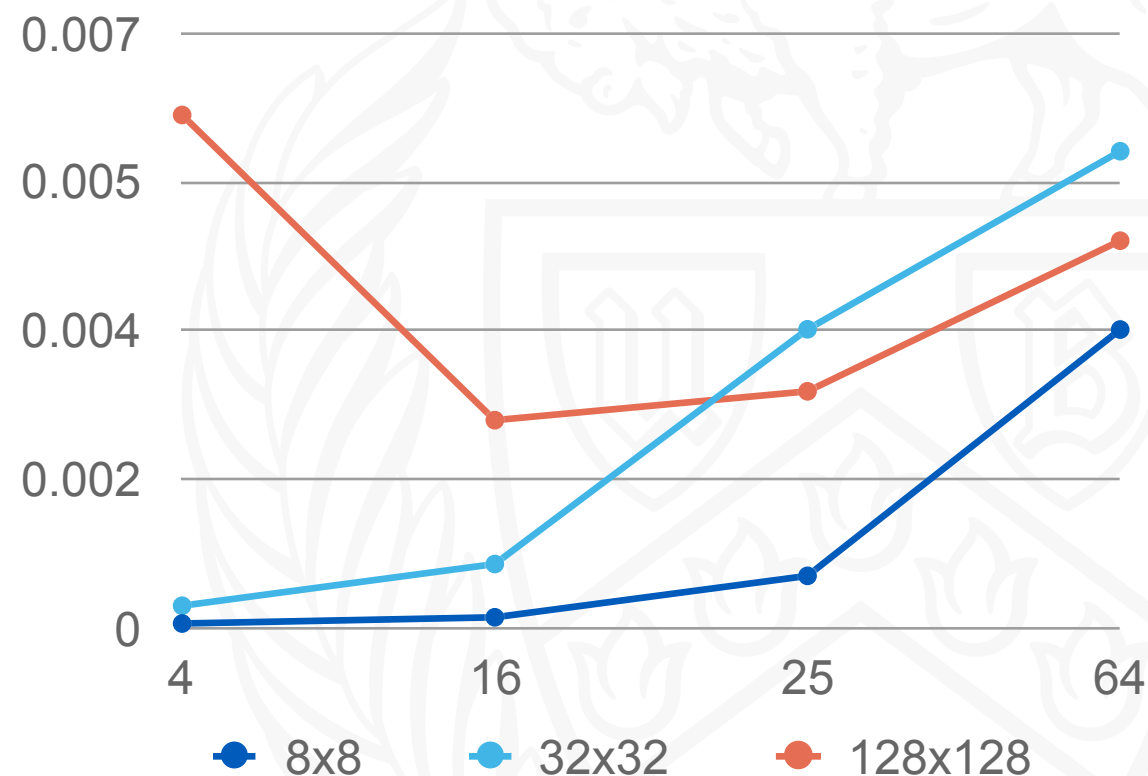
# Parallel Processing

	9	16	25	64
90x90	0.001354	0.001467	0.001426	0.003325
300x300	0.039349	0.022482	0.016006	0.008147
900x900	0.564426	0.415001	0.207663	0.104493
3000x3000	21.176418	12.546824	7.677214	3.003271
6000x6000	156.690799	92.761089	83.312945	27.089127



# Increasing the number of processors

	8x8	32x32	128x128	1000x1000
4	0.000061	0.000270	0.006036	1.924891
16	0.000134	0.000760	0.002450	0.830853
25	0.000621	0.003518	0.002789	0.190719
64	0.003515	0.005610	0.003559	0.057688



# Learnings

- Understanding MPI and Parallel processing.
- Cannon's Matrix Multiplication Algorithm
- Parallel processing and its effect on runtime.
- Understanding that just increasing the nodes won't always reduce the runtime.



## Next Steps

- Run single block matrix multiplication in parallel.
- Implementation using OpenMP.





**THANK YOU**

