**Information Technology** 

## FIT3143 - LECTURE WEEK 8

PARALLEL ALGORITHM DESIGN PARTITIONING BASED ON MATRIX OPERATIONS

algorithm distributed systems database systems computation knowledge madesign e-business model data mining interpretation distributed systems database software computation knowledge management and

## **Topic Overview**

- Matrix Algorithms & Problem Statement
- Decomposition
- Decomposition Fox's method

A portion of the content in the following slides were created by:

- a) Gergel V.P., Nizhni Novgorod, Introduction to Parallel Programming: Matrix Multiplication, 2005.
- b) Ananth Grama, Anshul Gupta, George Karypis, and Vipin Kumar, "Introduction to Parallel Computing", Addison Wesley, 2003.

ANANTH GRAMA • ANSHUL GUPTA GEORGE KARYPIS • VIPIN KUMAR Introduction to Parallel Computing **Second Edition** 

## **Matrix Algorithms: Introduction**

- Due to their regular structure, parallel computations involving matrices and vectors readily lend themselves to data-decomposition.
- Typical algorithms rely on input, output, or intermediate data decomposition.
- Most algorithms use one- and two-dimensional block, cyclic, and block-cyclic partitionings.

#### **Problem Statement**

#### Matrix multiplication:

$$C = A \cdot B$$

or

$$\begin{pmatrix} c_{0,0}, & c_{0,1}, & ..., & c_{0,l-1} \\ & & & & \\ c_{m-1,0}, & c_{m-1,1}, & ..., & c_{m-1,l-1} \end{pmatrix} = \begin{pmatrix} a_{0,0}, & a_{0,1}, & ..., & a_{0,n-1} \\ & & & & \\ a_{m-1,0}, & a_{m-1,1}, & ..., & a_{m-1,n-1} \end{pmatrix} \begin{pmatrix} b_{0,0}, & b_{0,1}, & ..., & a_{0,l-1} \\ & & & & \\ b_{n-1,0}, & b_{n-1,1}, & ..., & b_{n-1,l-1} \end{pmatrix}$$

The matrix multiplication problem can be reduced to the execution of m·l independent operations of matrix A rows and matrix B columns inner product calculation

$$c_{ij} = (a_i, b_j^T) = \sum_{k=0}^{n-1} a_{ik} \cdot b_{kj}, \ 0 \le i < m, \ 0 \le j < l$$

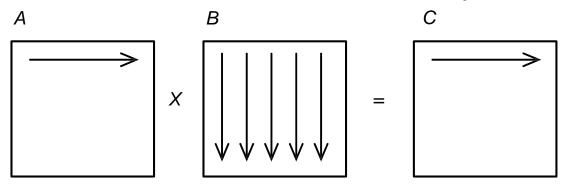
# Data parallelism can be exploited to design parallel computations

### Sequential Algorithm

```
// Sequential algorithm of matrix multiplication
double MatrixA[Size][Size];
double MatrixB[Size][Size];
double MatrixC[Size][Size];
int i,j,k;
for (i=0; i<Size; i++) {
  for (j=0; j<Size; j++) {
   MatrixC[i][j] = 0;
    for (k=0; k<Size; k++) {
      MatrixC[i][j] = MatrixC[i][j] + MatrixA[i][k]*MatrixB[k][j];
```

### Sequential Algorithm

- Algorithm performs the matrix C rows calculation sequentially
- At every iteration of the outer loop on i variable a single row of matrix A and all columns of matrix B are processed



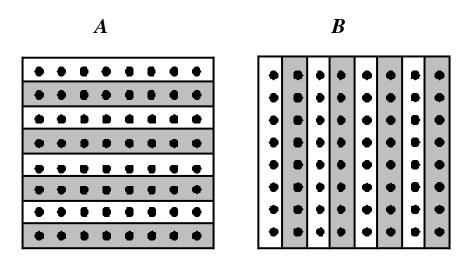
- m·l inner products are calculated to perform the matrix multiplication
- The complexity of the matrix multiplication is O(mnl).

 A fine-grained approach – the basic subtask is calculation of one element of matrix C

$$c_{ij} = (a_i, b_j^T), a_i = (a_{i0}, a_{i1}, ..., a_{in-1}), b_j^T = (b_{0j}, b_{1j}, ..., b_{n-1j})^T$$

- Number of basic subtasks is equal to n².
- As a rule, the number of available processors is less then n² (p<n²), so it will be necessary to perform the subtask scaling

- The aggregated subtask the calculation of one row of matrix C (the number of subtasks is n)
- Data distribution rowwise block-striped decomposition for matrix A and columnwise blockstriped decomposition for matrix B



#### Analysis of Information Dependencies

Each subtask hold one row of matrix **A** and one column of matrix **B**,

- At every iteration each subtask performs the inner product calculation of its row and column, as a result the corresponding element of matrix **C** is obtained
- Then every subtask i, 0≤ i<n, transmits its column of matrix</li>
   B for the subtask with the number (i+1) mod n.

After all algorithm iterations all the columns of matrix **B** were come within each subtask one after another

# Aggregating and Distributing the Subtasks among the Processors:

- In case when the number of processors p is less than the number of basic subtasks n, calculations can be aggregated in such a way that each processor would execute several inner products of matrix A rows and matrix B columns. In this case after the completion of computation, each aggregated basic subtask determines several rows of the result matrix C.
- Under such conditions the initial matrix A is decomposed into p
  horizontal stripes and matrix B is decomposed into p vertical
  stripes.
- Subtasks distribution among the processors have to meet the requirements of effective representation of the ring structure of subtask information dependencies.

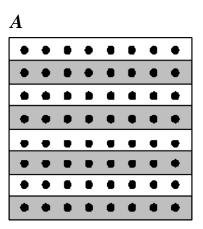
### Efficiency Analysis...

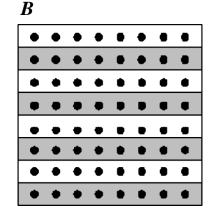
Speed-up and Efficiency generalized estimates

$$S_p = \frac{n^3}{(n^3/p)} = p$$
  $E_p = \frac{n^3}{p \cdot (n^3/p)} = 1$ 

Developed method of parallel computations allows to achieve ideal speed-up and efficiency characteristics

 Another possible approach for the data distribution is the rowwise block-striped decomposition for matrices A and B



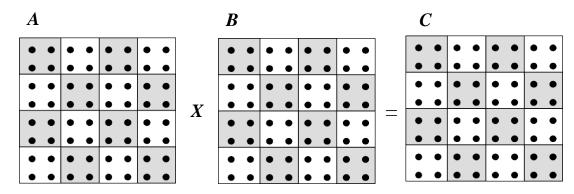


#### Analysis of Information Dependencies

- Each subtask hold one row of matrix **A** and one row of matrix **B**,
- At every iteration the subtasks perform the element-toelement multiplications of the rows; as a result the row of partial results for matrix **C** is obtained,
- Then every subtask i, 0≤ i<n, transmits its row of matrix B</li>
   for the subtask with the number (i+1) mod n.

After all algorithm iterations all rows of matrix **B** were come within every subtask one after another

#### **Data distribution** – checkerboard scheme



# **Basic subtask** is a procedure, that calculates all elements of one block of matrix **C**

$$\begin{pmatrix} A_{00}A_{01}...A_{0q-1} \\ ... \\ A_{q-10}A_{q-11}...A_{q-1q-1} \end{pmatrix} \times \begin{pmatrix} B_{00}B_{01}...B_{0q-1} \\ ... \\ B_{q-10}B_{q-11}...B_{q-1q-1} \end{pmatrix} = \begin{pmatrix} C_{00}C_{01}...C_{0q-1} \\ ... \\ c_{q-10}C_{q-11}...C_{q-1q-1} \end{pmatrix}, \quad C_{ij} = \sum_{s=0}^{q-1}A_{is}B_{sj}$$

#### Analysis of Information Dependencies

- Subtask with (i,j) number calculates the block C<sub>ij</sub>, of the result matrix C. As a result, the subtasks form the qxq twodimensional grid,
- Each subtask holds 4 matrix blocks:
  - block  $C_{ij}$  of the result matrix C, which is calculated in the subtask,
  - block  $\mathbf{A}_{ij}$  of matrix  $\mathbf{A}$ , which was placed in the subtask before the calculation starts,
  - blocks A<sub>ij</sub>' and B<sub>ij</sub>' of matrix A and matrix B, that are received by the subtask during calculations.

# **Analysis of Information Dependencies** – during iteration *I,* 0≤ *I*<*q,* algorithm performs:

The subtask (i,j) transmits its block A<sub>ij</sub> of matrix A to all subtasks of the same horizontal row i of the grid; the j index, which determines the position of the subtask in the row, can be obtained using equation:

$$j = (i+1) \mod q$$
,

where mod operation is the procedure of calculating the remainder of integer-valued division,

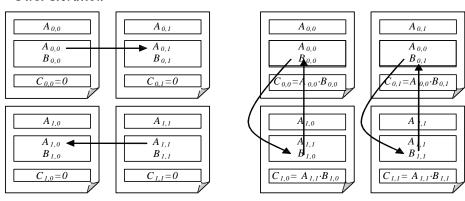
- Every subtask performs the multiplication of received blocks  $\mathbf{A}_{ij}$ ' and  $\mathbf{B}_{ij}$ ' and adds the result to the block  $\mathbf{C}_{ij}$ 

$$C_{ij} = C_{ij} + A'_{ij} \times B'_{ij}$$

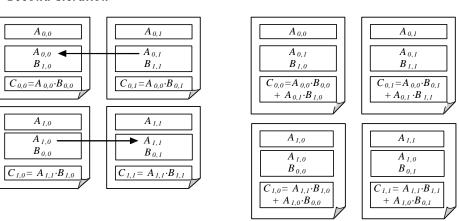
– Every subtask (i,j) transmits its block  $\mathbf{B}_{ij}$ ' to the neighbor, which is previous in the same vertical line (the blocks of subtasks of the first row are transmitted to the subtasks of the last row of the grid).

### Scheme of Information Dependences

#### First Iteration



#### Second Iteration



# Scaling and Distributing the Subtasks among the Processors

- The sizes of the matrices blocks can be selected so that the number of subtasks will coincides the number of available processors p,
- The most efficient execution of the parallel the Fox's algorithm can be provided when the communication network topology is a two-dimensional grid,
- In this case the subtasks can be distributed among the processors in a natural way: the subtask (i,j) has to be placed to the p<sub>i,j</sub> processor

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#### In depth discussion & example

- Please refer to the enclosed report attached with these slides, "Design and Implementation of Parallel Matrix Multiplication Algorithms using Message Passing Interface" by Chin-Kit Ng for further in-depth discussion and code examples.
  - Serial matrix multiplication example
  - Bernstein analysis for data dependency
  - Parallel matrix multiplication examples using POSIX and MPI