

#### 并行与分布式计算 Parallel & Distributed Computing

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#### **Lecture 9 — Programming with MPI**

- Matrix-Vector Multiplication and MPI + OpenMP

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#### **Outline:**

- > Example: matrix-vector multiplication
  - **□** Sequential algorithm;
  - **□** Design, analysis, and implementation of three parallel programs;
    - Rowwise block striped
    - Columnwise block striped
    - Checkerboard block (棋盘块)
- > MPI + OpenMP



#### Sequential Algorithm

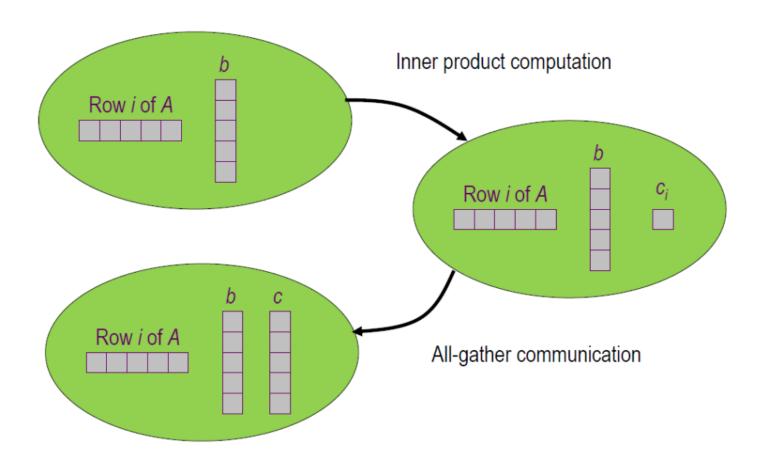
2	1	0	4		1		9
3	2	1	1		3	_	14
4	3	1	2	×	4	=	19
3	0	2	0		1		11



#### Rowwise Block Striped Matrix

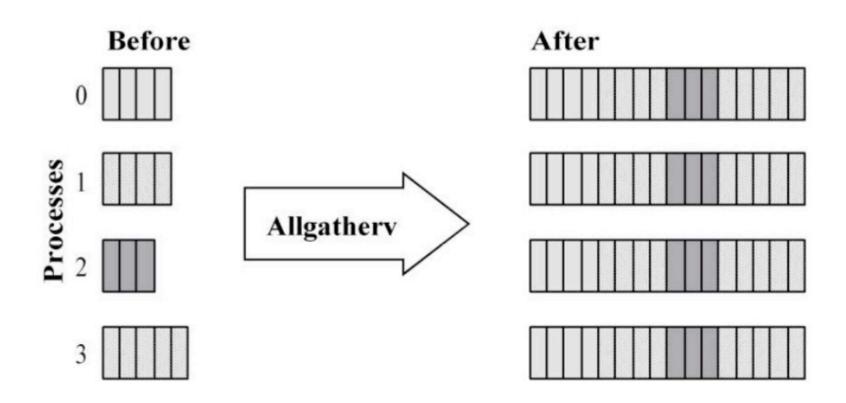
- > Partitioning through domain decomposition
- > Primitive task associated with
  - □ Row of matrix
  - **□** Entire vector

# Phases of Parallel Algorithm





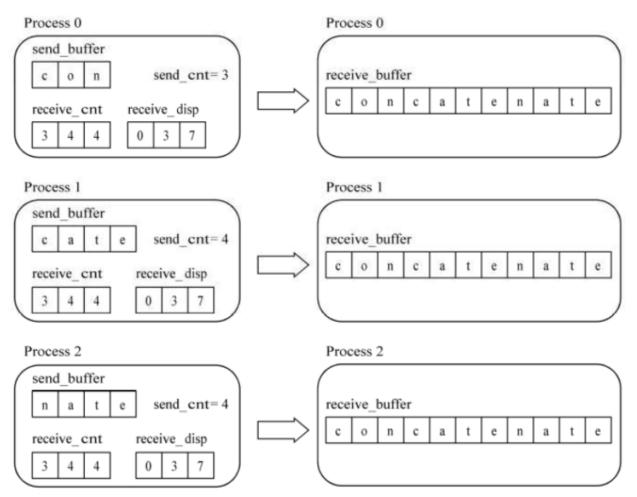
#### MPI\_Allgatherv



## MPI\_Allgatherv

```
int MPI Allgatherv (
void *send buffer,
int send cnt,
MPI Datatype send type,
void *receive buffer,
int *receive cnt,
int *receive disp,
MPI Datatype receive type,
MPI Comm communicator)
Use a simpler function MPI Allgather(...)
when the number of elements per processor
is a constant
```

#### MPI\_Allgatherv in Action





- > Static number of tasks
- > Regular communication pattern (all-gather)
- > Computation time per task is constant
- > Strategy:
  - **□** Agglomerate groups of rows
  - ☐ Create one task per MPI process



#### Columnwise Block Striped Matrix

- > Partitioning through domain decomposition
- > Task associated with
  - □ Column of matrix
  - **□** Vector element



#### Matrix-Vector Multiplication

$$c_{0} = a_{0,0} b_{0} + a_{0,1} b_{1} + a_{0,2} b_{2} + a_{0,3} b_{3} + a_{4,4} b_{4}$$

$$c_{1} = a_{1,0} b_{0} + a_{1,1} b_{1} + a_{1,2} b_{2} + a_{1,3} b_{3} + a_{1,4} b_{4}$$

$$c_{2} = a_{2,0} b_{0} + a_{2,1} b_{1} + a_{2,2} b_{2} + a_{2,3} b_{3} + a_{2,4} b_{4}$$

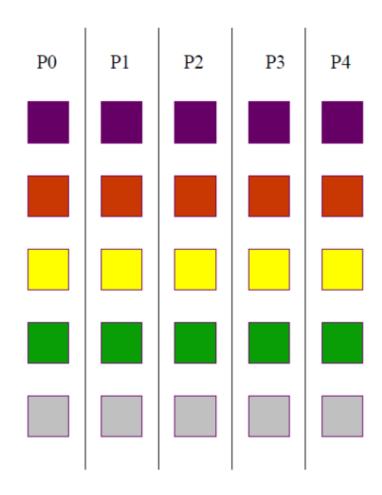
$$c_{3} = a_{3,0} b_{0} + a_{3,1} b_{1} + a_{3,2} b_{2} + a_{3,3} b_{3} + b_{3,4} b_{4}$$

$$c_{4} = a_{4,0} b_{0} + a_{4,1} b_{1} + a_{4,2} b_{2} + a_{4,3} b_{3} + a_{4,4} b_{4}$$
Proc 4

Processor 1's initial computation

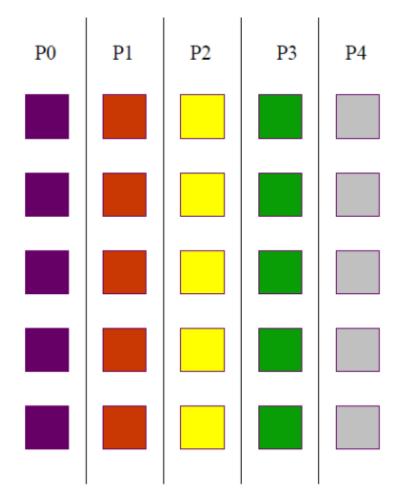
Processor 0's initial computation

#### All-to-all Exchange (Before)





#### All-to-all Exchange (After)

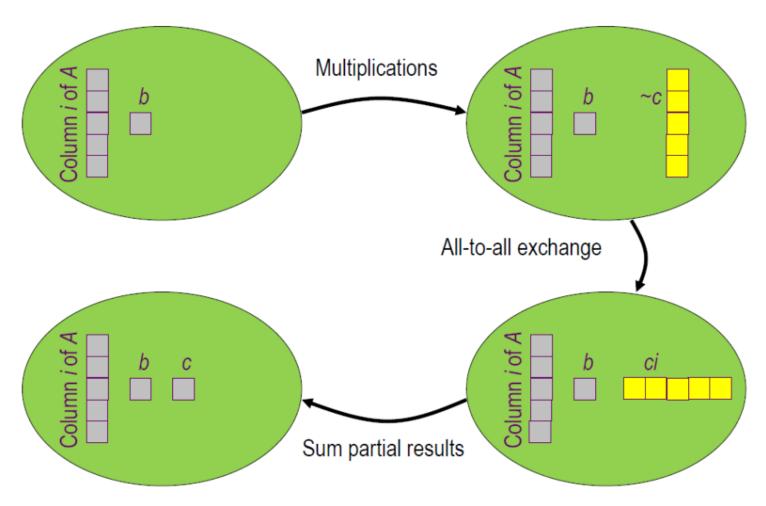


# All-to-All Exchange (All-to-All Personalized Communication)



How to improve the communication efficiency?

#### Phases of Parallel Algorithm



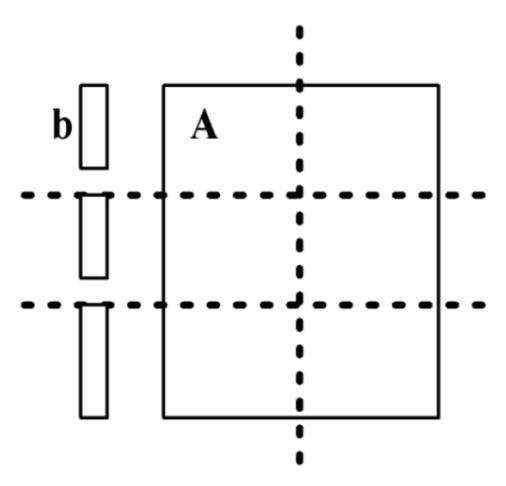


- > Static number of tasks
- > Regular communication pattern (all-to-all)
- > Computation time per task is constant
- > Strategy:
  - **□** Agglomerate groups of columns
  - ☐ Create one task per MPI process

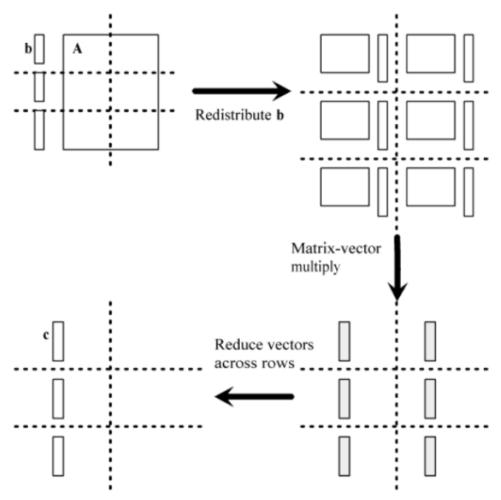


- $\triangleright$  Associate primitive task with each element of the matrix A
- > Each primitive task performs one multiply
- > Agglomerate primitive tasks into rectangular blocks
- > Processes form a 2-D grid
- $\triangleright$  Vector b distributed by blocks among processes in first column of grid

## Tasks after Agglomeration



#### Algorithm's Phases

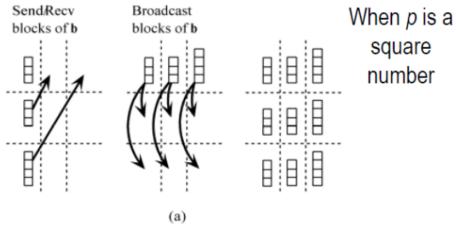


#### Redistributing Vector b

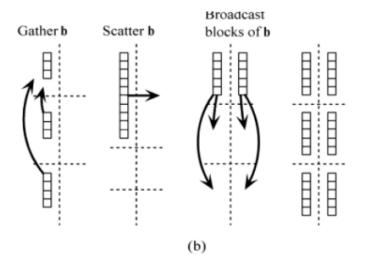
- $\triangleright$  Step 1: Move b from processes in first row to processes in first column
  - $\square$  If p square
    - First column/first row processes send/receive portions of b
  - $\square$  If p not square
    - Gather b on process 0, 0
    - Process 0, 0 broadcasts to first row procs
- $\triangleright$  Step 2: First row processes scatter b within columns



#### Algorithm's Phases



When p is not a square number





- Want processes in a virtual 2-D grid
- > Create a custom communicator to do this
- > Collective communications involve all processes in a communicator
- ➤ We need to do broadcasts, reductions among subsets of processes
- > We will create communicators for processes in same row or same column



- > Process group
- > Context
- > Attributes
  - Topology (let's address processes another way)
  - Others we won't consider



- > MPI\_Dims\_create
  - **□** Input parameters
    - Total number of processes in desired grid
    - Number of grid dimensions
  - **□** Returns number of processes in each dim
- > MPI Cart create
  - □ Creates communicator with Cartesian topology (笛卡尔拓扑结构)

#### MPI\_Dims\_create

dims	function call	dims
before call		on return
(0,0)	MPI_DIMS_CREATE(6, 2, dims)	(3,2)
(0,0)	MPI_DIMS_CREATE(7, 2, dims)	(7,1)
(0,3,0)	MPI_DIMS_CREATE(6, 3, dims)	(2,3,1)
(0,3,0)	MPI_DIMS_CREATE(7, 3, dims)	erroneous call

source: DeinoMPI

# MPI\_Cart\_create (创建笛卡尔拓扑)

```
int MPI Cart create (
  MPI Comm old comm, /* Input - old communicator */
   int dims, /* Input - grid dimensions */
   int *size, /* Input - # procs in each dim */
   int *periodic,
       /* Input - periodic[j] is 1 if dimension j
          wraps around; 0 otherwise */
   int reorder,
       /* 1 if process ranks can be reordered */
  MPI Comm *cart comm)
      /* Output - new communicator */
```

#### **Using** MPI\_Dims\_create **and** MPI\_Cart\_create

```
MPI Comm cart comm;
int p;
int periodic[2];
int size[2];
size[0] = size[1] = 0;
MPI Dims create (p, 2, size);
periodic[0] = periodic[1] = 0;
MPI Cart create (MPI COMM WORLD, 2, size,
1, &cart comm);
```

#### **Using** MPI\_Dims\_create **and** MPI\_Cart\_create

```
MPI Comm cart comm;
int p;
int periodic[2];
int size[2];
size[0] = size[1] = 0;
MPI Dims create (p, 2, size);
periodic[0] = periodic[1] = 0;
MPI Cart create (MPI COMM WORLD, 2, size,
1, &cart comm);
```



- > MPI\_Cart\_rank
  - ☐ Given coordinates of process in Cartesian communicator, returns process rank
- > MPI Cart cords
  - ☐ Given rank of process in Cartesian communicator, returns process' coordinates

## **Header for** MPI\_Cart\_rank

```
int MPI Cart rank (
   MPI Comm comm,
      /* In - Communicator */
   int *coords,
      /* In - Array containing process'
              grid location */
   int *rank)
      /* Out - Rank of process at
         specified coords */
```

#### **Header for** MPI\_Cart\_coords

```
int MPI Cart coords (
   MPI Comm comm,
      /* In - Communicator */
   int rank,
      /* In - Rank of process */
   int dims,
      /* In - Dimensions in virtual grid
* /
   int *coords)
      /* Out - Coordinates of specified
         process in virtual grid */
```



#### **Collective Communications**

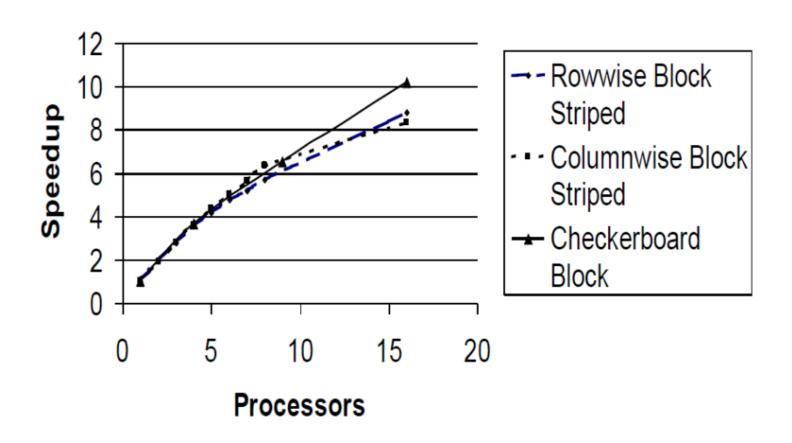
	Ring	2-D Mesh	Hypercube
One-to-All Broadcast			
All-to-One Reduction			
All-to-All Broadcast			
All-to-All Reduction			
Scatter			
Gather			
All-to-All Personalized Communication			

#### 重点和精华



Operation	MPI Name	
One-to-all broadcast	MPI_Bcast	
All-to-one reduction	MPI_Reduce	
All-to-all broadcast	MPI_Allgather	
All-to-all reduction	MPI_Reduce_scatter	
All-reduce	MPI_Allreduce	
Gather	MPI_Gather	
Scatter	MPI_Scatter	
All-to-all personalized	MPI_Alltoall	

#### Comparison of Three Algorithms







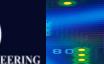
**Programming with MPI** 

MPI + OpenMP



#### MPI + OpenMP

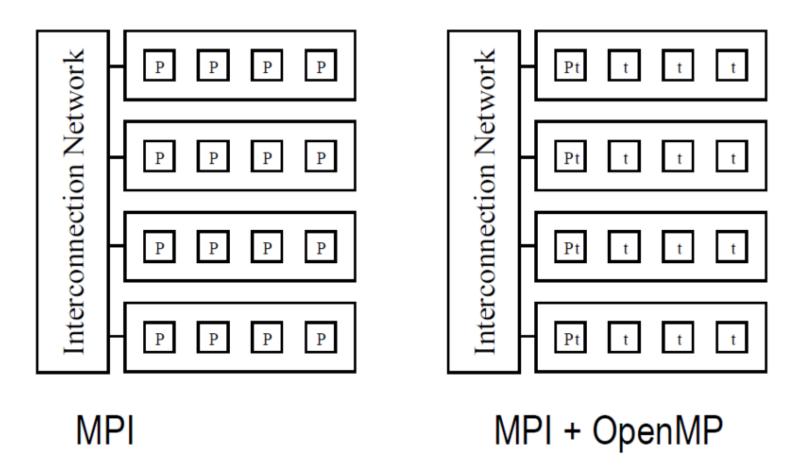
- > Advantages of using both MPI and OpenMP
  - ☐ Case Study: Jacobi method
    - An iterative methods to solve linear systems







# MPI vs. MPI + OpenMP



Michael J.Quinn, "Parallel Programming in C with MPI and OpenMP," 2003.



- > Lower communication overhead
  - $\square$  Message passing with mk processes, versus
  - $\square$  Message passing with m processes with k threads each
- > More portions of program may be practical to parallelize
- > May allow more overlap of communications with computations
  - ☐ For example, if 3 MPI processes are waiting for messages, and 1

MPI process is active, it is worthwhile to fork some threads to

speedup the 4th process

Michael J. Quinn, "Parallel Programming in C with MPI and OpenMP," 2003.



- > For example, a serial program runs in 100s
  - ☐ S: 5s is inherent sequential
  - □ P<sub>1</sub>: 5s are parallelizable but not worth message passing
  - □ P<sub>2</sub>: 90s are perfectly parallelizable
- > MPI-only with 16 processes
  - □ Replicate the P<sub>1</sub> part of program on all processes
  - $\square$  Speedup = 1 / (0.10 + 0.90/16) = 6.4
- > Hybrid MPI & threads
  - **■** Execute the replicated P<sub>1</sub> with 2 threads
  - $\square$  Speedup = 1 / (0.05 + 0.05/2 + 0.90/16) = 7.6
  - **□** 19% faster than MPI-only

Michael J.Quinn, "Parallel Programming in C with MPI and OpenMP," 2003.

# Code for matrix\_vector\_product

```
void matrix vector product (int id, int p,
   int n, double **a, double *b, double *c)
   int i, j;
              /* Accumulates sum */
   double tmp;
   for (i=0; i<BLOCK SIZE(id,p,n); i++) {
      tmp = 0.0;
      for (j = 0; j < n; j++)
        tmp += a[i][j] * b[j];
     piece[i] = tmp;
   new replicate block vector (id, p,
     piece, n, (void *) c, MPI DOUBLE);
```

# Adding OpenMP Directives

- > Want to minimize fork/join overhead by making parallel the outermost possible loop
- ➤ Outer loop may be executed in parallel if each thread has a private copy of tmp and j

```
#pragma omp parallel for private(j,tmp)
for (i=0; i<BLOCK_SIZE(id,p,n); i++) {</pre>
```



#### **User Control of Threads**

- > Want to give user opportunity to specify number of active threads per process
- > Add a call to omp set num threads to function main
- > Argument comes from command line



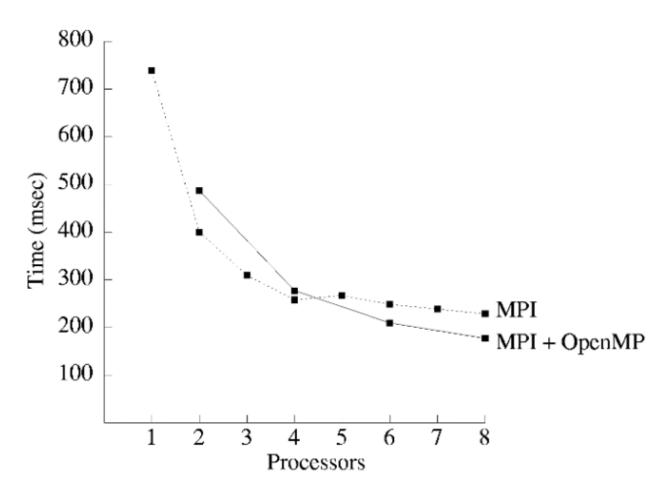
# What Happened?

> We transformed a MPI program to a MPI+OpenMP program by adding only two lines to our program!

#### Benchmarking

- > Target system: a commodity cluster with four dual processor nodes
- > MPI program executes on 1, 2, ..., 8 CPUs
  - □ On 1, 2, 3, 4 CPUs, each process on different node, maximizing memory bandwidth per CPU
- ➤ MPI+OpenMP program executes on 1, 2, 3, 4 processes
  - **□** Each process has two threads
  - □ C+MPI+OpenMP program executes on 2, 4, 6, 8 threads

#### Results of Benchmarking



Michael J.Quinn, "Parallel Programming in C with MPI and OpenMP," 2003.

# Analysis of Results

- ➤ MPI+OpenMP program slower on 2, 4 CPUs because MPI+OpenMP threads are sharing memory bandwidth, while C+MPI processes are not
- > MPI+OpenMP programs faster on 6, 8 CPUs because they have lower communication cost

#### Case Study: Jacobi Method

- > Begin with MPI program that uses Jacobi method to solve steady state heat distribution problem
- ➤ Program based on rowwise block striped decomposition of two-dimensional matrix containing finite difference mesh (有限差分网格, FDM)



# Gaussian Elimination (高斯消元)

- $\triangleright$  Used to solve Ax = b when A is dense
- ightharpoonup Reduces Ax = b to upper triangular system Tx = c
- $\triangleright$  Back substitution can then solve Tx = c for x



# Gaussian Elimination Example (1/4)

$$4x_0 +6x_1 +2x_2 -2x_3 = 8$$

$$2x_0 +5x_2 -2x_3 = 4$$

$$-4x_0 - 3x_1 - 5x_2 + 4x_3 = 1$$

$$8x_0 + 18x_1 - 2x_2 + 3x_3 = 40$$



# Gaussian Elimination Example (2/4)

$$4x_0 +6x_1 +2x_2 -2x_3 = 8$$

$$-3x_1 +4x_2 -1x_3 = 0$$

$$+3x_1 -3x_2 +2x_3 = 9$$

$$+6x_1 -6x_2 +7x_3 = 24$$

Michael J. Quinn, "Parallel Programming in C with MPI and OpenMP," 2003.



# Gaussian Elimination Example (3/4)

$$4x_0 +6x_1 +2x_2 -2x_3 = 8$$

$$-3x_1 +4x_2 -1x_3 = 0$$

$$1x_2 + 1x_3 = 9$$

$$2x_2 + 5x_3 = 24$$



# Gaussian Elimination Example (4/4)

$$4x_0 +6x_1 +2x_2 -2x_3 = 8$$

$$-3x_1 + 4x_2 - 1x_3 = 0$$

$$1x_2 + 1x_3 = 9$$

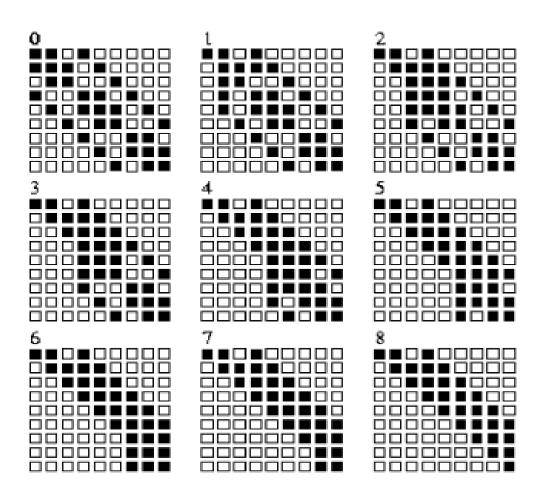
$$3x_3 = 6$$



# Sparse Systems

- > Gaussian elimination not well-suited for sparse systems
- > Coefficient matrix gradually fills with nonzero elements
  - **□** Increases storage requirements
  - **□** Increases total operation count

#### Example of "Fill" in Gaussian Elimination



Michael J.Quinn, "Parallel Programming in C with MPI and OpenMP," 2003.



#### Iterative Methods

- > Iterative method: algorithm that generates a series of approximations to solution's value
- > Require less storage than direct methods
- > Since they avoid computations on zero elements, they can save a lot of computations



#### Jacobi Method

Let

$$A\mathbf{x} = \mathbf{b}$$

be a square system of *n* linear equations, where:

$$A = egin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \ a_{21} & a_{22} & \cdots & a_{2n} \ dots & dots & \ddots & dots \ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}, \qquad \mathbf{x} = egin{bmatrix} x_1 \ x_2 \ dots \ x_n \end{bmatrix}, \qquad \mathbf{b} = egin{bmatrix} b_1 \ b_2 \ dots \ b_n \end{bmatrix}.$$

Then A can be decomposed into a diagonal component D, and the remainder R:

$$A = D + R \qquad ext{where} \qquad D = egin{bmatrix} a_{11} & 0 & \cdots & 0 \ 0 & a_{22} & \cdots & 0 \ dots & dots & \ddots & dots \ 0 & 0 & \cdots & a_{nn} \end{bmatrix} ext{and } R = egin{bmatrix} 0 & a_{12} & \cdots & a_{1n} \ a_{21} & 0 & \cdots & a_{2n} \ dots & dots & \ddots & dots \ a_{n1} & a_{n2} & \cdots & 0 \end{bmatrix}.$$

The solution is then obtained iteratively via

$$\mathbf{x}^{(k+1)} = D^{-1}(\mathbf{b} - R\mathbf{x}^{(k)}),$$

where  $\mathbf{x}^{(k)}$  is the kth approximation or iteration of  $\mathbf{x}$  and  $\mathbf{x}^{(k+1)}$  is the next or k+1 iteration of  $\mathbf{x}$ . The element-based formula is thus:

$$x_i^{(k+1)} = rac{1}{a_{ii}}\left(b_i - \sum_{j 
eq i} a_{ij}x_j^{(k)}
ight), \quad i=1,2,\ldots,n.$$

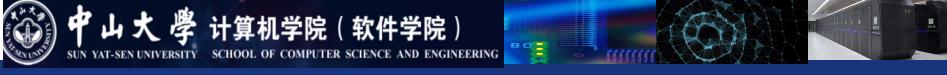
Values of elements of vector x at iteration k+1 depend upon values of vector x at iteration k

Michael J. Quinn, "Parallel Programming in C with MPI and OpenMP," 2003.



# Rate of Convergence

- > Even when Jacobi method, rate of convergence often too slow to make them practical
- > We will move on to an iterative method with much faster convergence

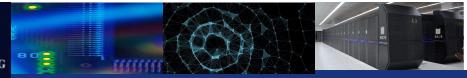


#### Methodology

- > Profile execution of MPI program
- Focus on adding OpenMP directives to most compute-intensive **function**



- ➤ Hybrid C+MPI+OpenMP program uniformly faster than C+MPI program
- > Computation/communication ratio of hybrid program is superior
- > Number of mesh points per element communicated is twice as high per node for the hybrid program
- ➤ Lower communication overhead leads to 19% better speedup on 8 CPUs



# Thank You!