## **Lecture 17: Array Algorithms**

## CS178: Programming Parallel and Distributed Systems

April 4, 2001 Steven P. Reiss

I. Overview

### A. We talking about constructing parallel programs

- 1. Last time we discussed sorting algorithms
- 2. Looking at various techniques

## **B.** The emphasis here shouldn't be on sorting per se

- 1. But rather on the underlying techniques
- 2. Divide and conquer (split the problem)
- 3. Finding algorithms that allow maximal parallelism
  - a) This is what bitonic merge does for us
- 4. Treating blocks as units

## C. Today I want to look at additional techniques

- 1. Numerical (array) problems
- 2. Many of the same techniques in a new context
- 3. New techniques -- pipelining

## **II. Matrix Computations**

## A. Motivation

### 1. Many scientific problems involve matrices

- a) We've seen one example with the heat distribution example we did before break
  - (1) Value at one point is related to the value at other points
  - (2) This generalizes to multiple values
- b) The homework has another example
  - (1) Maxtrix to store responses
  - (2) Need to compute over that
- c) Other problems can be cast as matrix problems

- (1) Web searching
- (2) Image processing

### 2. Many of these problems can be large

- a) Matrices of  $10^5 * 10^5 * 10^5$  for weather, flow, ...
- b) Number of web pages
- c) Number of users
- 3. If we want to do these practically, we need to do them in parallel

## **B.** What are the basic problems

### 1. Matrix multiplication

- a) Occurs in many problems
- b) Part of many other problems (transitive closure)
- c) Same algorithm as many other problems
- d) Matrix-vector multiplication is a common subset
  - (1) Lots of vectors done at once
- e) Recall the standard sequential algorithm

### 2. Gaussian elimination

- a) Solving systems of linear equations
- b) We've seen instances of this already
  - (1) Heat distribution problem with sparse matrices
  - (2) This was solved using iterative methods
  - (3) These don't work as well for full matrices
- c) Related to matrix inversion

## C. How is the array stored

- 1. This is going to depend on the problem
- 2. But also is going to control the algorithm
- 3. Alternatives
  - a) All in one node
  - b) All in all nodes
  - c) On disk (to be read by one or more nodes)
  - d) Spread out over all the nodes (e.g. heat flow problem)

## **III.Matrix Multiplication**

## A. Obvious ways of parallelizing

### 1. The inner loop can be done in parallel

### 2. Actually any of the loops can be

a) This is a standard technique

- 3. But this means that each node needs the whole array
- 4. This is what is done with parallel fortran, etc.
- 5. This corresponds to row or column orientation
  - a) We'll get back to this

### B. Can break the matrix up into blocks

### 1. Note that this works mathematically

a) Block sizes need to be compatible x\*y and y\*z

### 2. This is again a standard technique

- a) We used it with sorting
- 3. But even so, how do we multiply

## **C. Recursive implementation**

- 1. Break the matrix in 4 blocks
- 2. If block is one element, then just multiply
- 3. Else
  - a) Compute the 8 cross products recursively
  - b) Add the submatrices to get the results

### 4. 8 recursive calls can be done in parallel

- a) Can be repeated until you run out of processors
- b) Seems to imply that the whole matrix is available or storable

### 5. Message based approaches shouldn't assume this

## D. Cannon's algorithm

### 1. Assume a wraparound mesh topology

- a) Assume there are  $\mathsf{P}^2$  processors  $\mathsf{P}_{i,j}$  holding submatrices (or elements)
- b) Initially  $P_{i,j}$  holds  $a_{i,j}$  and  $b_{i,j}$

c) We then shift rows and columns around the matrix to do the multiplication

### 2. Algorithm

- a) Initialize: move items to "aligned" position
  - Move Row i of A i places left Move Column j of B j places upward
  - (1) Then  $P_{i,j}$  contains  $a_{i,j+i}$  and  $b_{i+j,j}$
  - (2) This is a part of the sum
- b) Do the initial multiplication
  - c = a\*b for each processor
  - (1) c, a, b represent the elements (matrices) held by that processor at that time
- c) Shift row i of A one place left; shift row j of B one place up
  - (1) This gives the next component of the sum
- d) Accumulate the new result c += a\*b
- e) Repeat c) and d) n-1 times to get the final result

## E. Fox's algorithm

## 1. As an alternative to moving whole rows and columns, we can broadcast elements

### 2. Code for Processor P<sub>i,j</sub>

```
dest = [i-1 mod n, j]
src = [i+1 mod n, j]
for (stage = 0; stage < n; ++stage) {
    kbar = (i+stage) mod n
    Broadcast A[i,kbar] across process row i
    C[i,j] += A[i,kbar]*B[kbar,j]
    Send B[kbar,j] to dest
    Receive B[kbar+1 mod n,j] from source
}</pre>
```

## F. Pipelined processing

- 1. Another way of doing the computation is to pipeline the processing
  - a) This is another example of a general technique
  - b) Program has send compute cycles
  - c) All processor operate in sync on those cycles
- 2. Basic idea:

- a) Send  $a_{0,0}$ ,  $a_{0,1}$ ,  $a_{0,2}$ , ... to first row, one per cycle
- b) Send  $b_{0,0}$ ,  $b_{0,1}$ ,  $b_{0,2}$ , ... to first column, one per cycle
- c) Send 0,  $a_{1,0}$ ,  $a_{1,1}$ ,  $a_{1,2}$ , to second row; etc
- d) At each step, processor computes product of its inputs and accumulates
- e) After 2N steps, everything is done
- 3. Note how this works
- 4. Note this can be applied to matrix-vector processing as well

## **IV. Systems of Linear Equations**

## A. Gaussian Elimination

### 1. Recall the sequential algorithm

- a) Set the diagonal to ones
- b) Set everything below the diagonal to zeros
- c) At each stage:
  - (1) Compute pivot (why and how)
  - (2) Compute multiplier m for each row  $A_{j,i} / A_{i,i}$
  - (3) Subtract row i \* m from row j
  - (4) Do the same for B<sub>i</sub>
- d) Back substitute at the end

### 2. How might you parallelize this (what techniques)

### 3. Partitioning

- a) Blocks don't work
- b) Want to partition into rows (or sets of rows)

## **B.** Pipelining

### 1. First row is broadcast to all other processors

- a) Each computes multiplier and then updates its row
- 2. Then second row is broadcast to remaining processors
  - a) Etc.
  - b) Note that this can be pipelined
    - (1) Processor gets data

(2) Processor resends data; computes; sends result

### **C.** Partitioning

- 1. This assumed one processor per row
- 2. What happens if we have fewer (more typical)
- 3. Blocks of rows (strips)
  - a) Here processors become idle

## 4. Cyclic partitioning

a) Assign rows sequentially to processors

## **D.** What about pivoting

### 1. This can be done by finding the pivot row

- a) Finding max element over processors, return index
- b) Then swapping the two rows

## 2. It can also be done by maintaining an index array

- a) This effectively swaps the rows -- index tells row number
- b) But this information needs to be shared
- c) This makes back substitution more difficult

# 3. Can imagine other approaches that pass rows thru a mesh

- a) Pass original row down to proper position
- b) Pass pivot row up and down to proper positions
- c) Compute as you get it

## V. Gettting the data to the nodes

# A. We assume that the nodes hold portions of the array

### 1. How do these portions get there in the first place

- a) Could be computed there
- b) Often, however, data must be read from a file or computed centrally

### 2. File I/O

- a) Don't want all nodes reading from same file (why?)
- b) Separate files means knowing configuration in advance
- 3. Assume one node reads file (or has data initially)

a) How to send it out

## **B. MPI Has scatter/gather facilities for this**

MPI\_Scatter(void \* sendptr,int sendcnt,MPI\_Datatype sendtype, void \* recvptr,int recvcnt,MPI\_Datatype recvtype, int root,MPI\_Comm comm) MPI\_Gather(void \* sendptr,int sendcnt,MPI\_Datatype sendtype, void \* recvptr,int recvcnt,MPI\_Datatype recvtype, int root,MPI\_Comm comm)

#### 1. Gather

- a) Each process sends info from its send area to the root
- b) Root receives the data and stores it in rank order
- c) Note that root receive area needs to be big enough

### 2. Scatter does the opposite

- a) Sends from root to all processors
- 3. This can be used for reading and distributing array
- **4.** Suppose we want all nodes to end up with the array MPI\_Allgather(...) [no root argument]

## C. MPI also has facilities for broadcasting

- 1. This is a send-recv type call
- 2. Root is doing the send, all others are receive
- 3. At the end, all will have the message in their buffer