# **Collective Communication**

#### John Mellor-Crummey

Department of Computer Science Rice University

johnmc@rice.edu



### **Group Communication**

- Motivation: accelerate interaction patterns within a group
- Approach: collective communication
  - —group works together *collectively* to realize a communication —constructed from pairwise point-to-point communications
- Implementation strategy
  - ----standard library of common collective operations
  - —leverage target architecture for efficient implementation
- Benefits of standard library implementations
  - -reduce development effort and cost for parallel programs
  - —improve performance through efficient implementations
  - —improve quality of scientific applications

### **Topics for Today**

- One-to-all broadcast and all-to-one reduction
- All-to-all broadcast and reduction
- All-reduce and prefix-sum operations
- Scatter and gather
- All-to-all personalized communication
- Optimizing collective patterns

### **Assumptions**

- Network is bidirectional
- Communication is single-ported

- Communication cost model
  - —message of size *m*, no congestion, time =  $t_s + t_w m$

—congestion: model by scaling  $t_w$ 

#### **One-to-All and All-to-One**

One-to-all broadcast

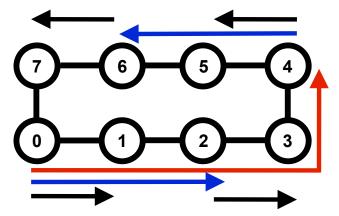
—a processor has *M* units of data that it must send to everyone

- All-to-one reduction
  - -each processor has *M* units of data
  - -data items must be combined using some associative operator
    - e.g. addition, min, max
  - -result must be available at a target processor

### **One-to-All and All-to-One on a Ring**

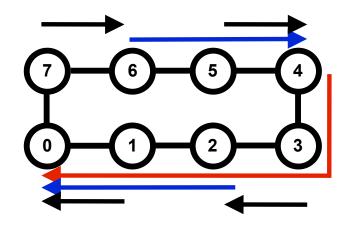
- Broadcast
  - -naïve solution
    - source sends send p 1 messages to the other p 1 processors
  - -use recursive doubling
    - source sends a message to a selected processor

yields two independent problems over halves of the machine



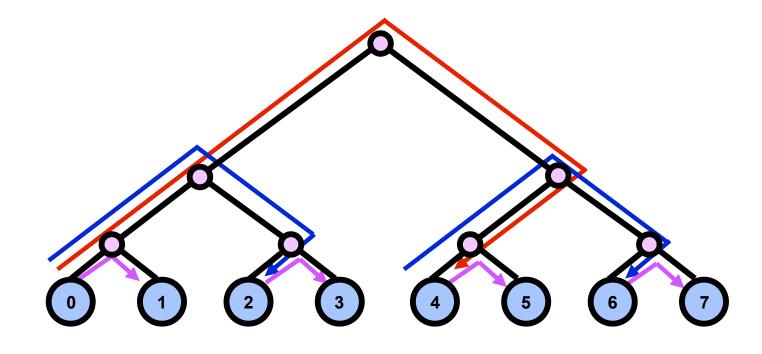
Reduction

— invert the process



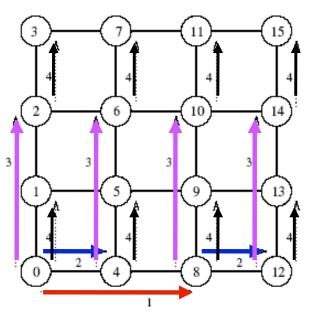
#### **Broadcast on a Balanced Binary Tree**

- Consider processors arranged in a dynamic binary tree
  - —processors are at the leaves
  - —interior nodes are switches
- Assume leftmost processor is the root of the broadcast
- Use recursive doubling strategy: log *p* stages



#### **Broadcast and Reduction on a 2D Mesh**

- Consider a square mesh of p nodes
  - treat each row as a linear array of  $p^{1/2}$  nodes
  - treat each column as a linear array of  $p^{1/2}$  nodes
- Two step broadcast and reduction operations
  - 1. perform the operation along a row
  - 2. perform the operation along each column concurrently



broadcast on 4 x 4 mesh

Generalizes to higher dimensional meshes

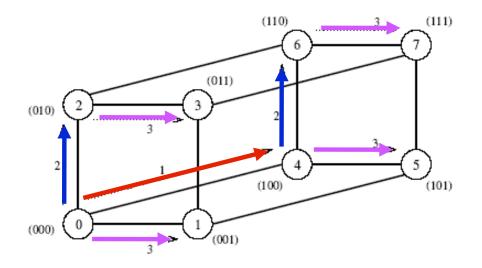
### **Broadcast and Reduction on a Hypercube**

• Consider hypercube with 2<sup>d</sup> nodes

-view as *d*-dimensional mesh with two nodes in each dimension

• Apply mesh algorithm to a hypercube

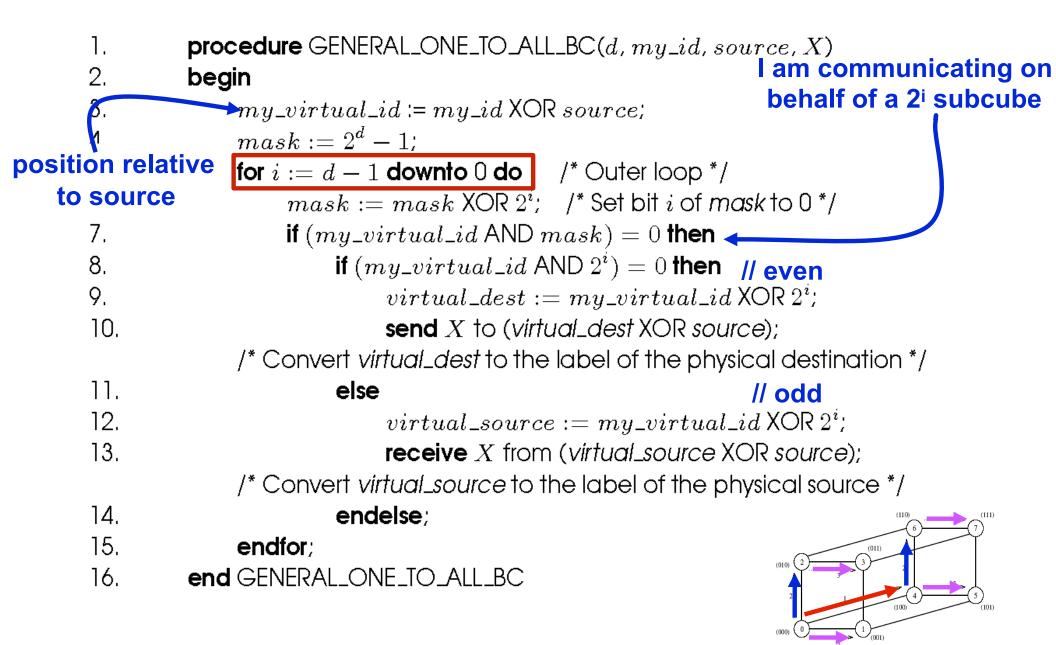
*—d* (= *log p*) steps



### **Broadcast and Reduction Algorithms**

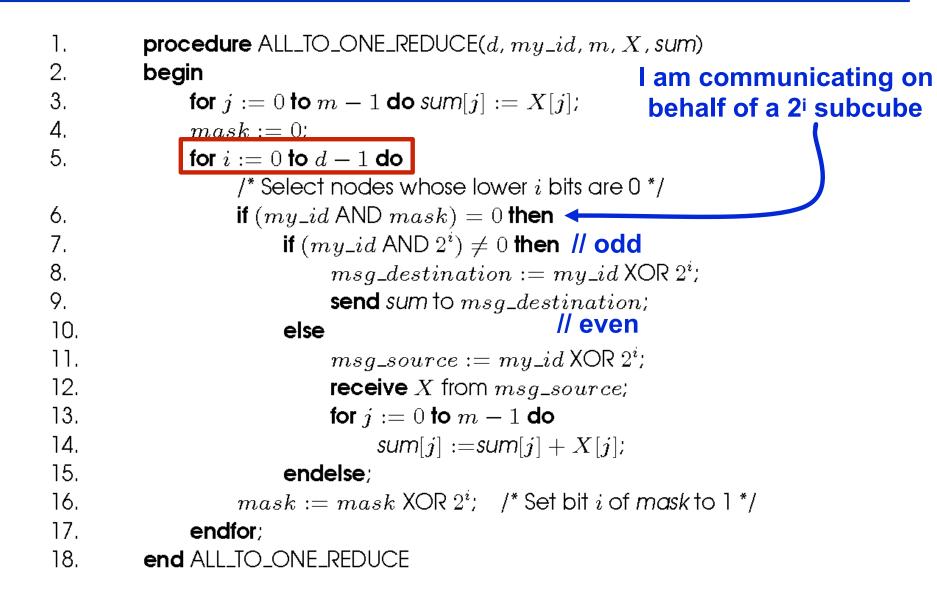
- Each of aforementioned broadcast/reduction algorithms —adaptation of the same algorithmic template
- Next slide: a broadcast algorithm for a hypercube of 2<sup>d</sup> nodes
  - -can be adapted to other architectures
  - —in the following algorithm
    - my\_id is the label for a node
    - X is the message to be broadcast

### **One-to-All Broadcast Algorithm**



**One-to-all broadcast of a message X from source on a hypercube** 11

#### **All-to-One Reduction Algorithm**



#### All-to-One sum reduction on a *d*-dimensional hypercube

Each node contributes msg X containing m words, and node 0 is the destination 12

### **Broadcast/Reduction Cost Analysis**

#### Hypercube

- Log p point-to-point simple message transfers
   —each message transfer time: t<sub>s</sub> + t<sub>w</sub>m
- Total time

$$T = (t_s + t_w m) \log p.$$

#### **All-to-All Broadcast and Reduction**

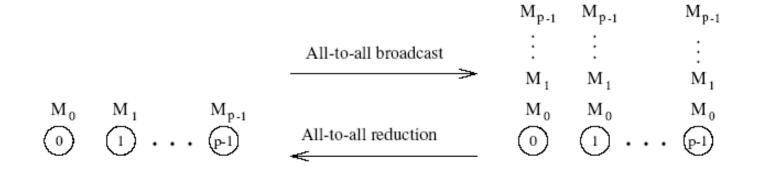
Each processor is the source as well as destination

• Broadcast

—each process broadcasts its own *m*-word message all others

Reduction

-each process gets a copy of the result

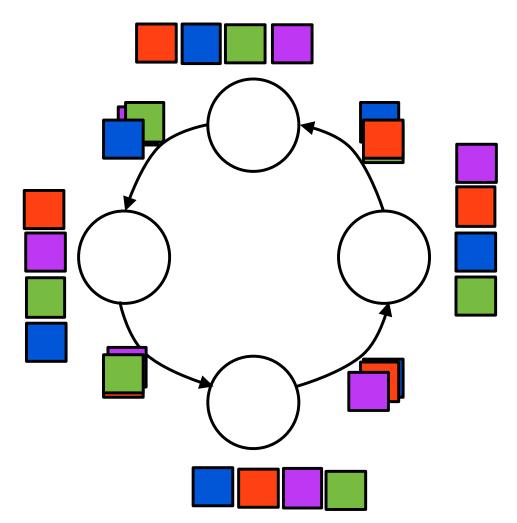


### **All-to-All Broadcast/Reduction on a Ring**

1.	<b>procedure</b> ALL_TO_ALL_BC_RING( <i>my_id</i> , <i>my_msg</i> , <i>p</i> , <i>result</i> )
2.	begin
3.	left := $(my\_id - 1) \mod p$ ;
4.	$right := (my_id + 1) \mod p;$
5.	result := my_msg;
6.	msg := result; <b>message size</b>
7.	for $i := 1$ to $p - 1$ do stays constant
8.	send msg to right;
9.	<b>receive</b> <i>msg</i> from <i>left</i> ;
10.	result := result $\cup$ msg;
11.	endfor;
12.	end ALL_TO_ALL_BC_RING

Also works for a linear array with bidirectional communication channels

#### **All-to-All Broadcast on a Ring**



For an all-to-all reduction

- combine (rather than append) each incoming message into your local result
- at each step, forward your incoming msg to your successor

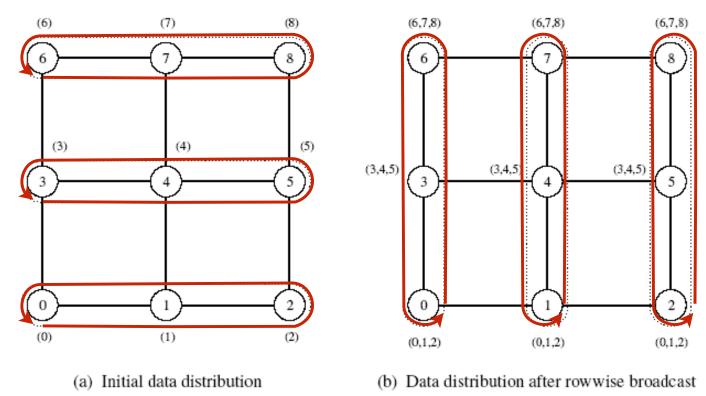
#### **All-to-all Broadcast on a Mesh**

#### **Two phases**

• Perform row-wise all-to-all broadcast as for linear array/ring

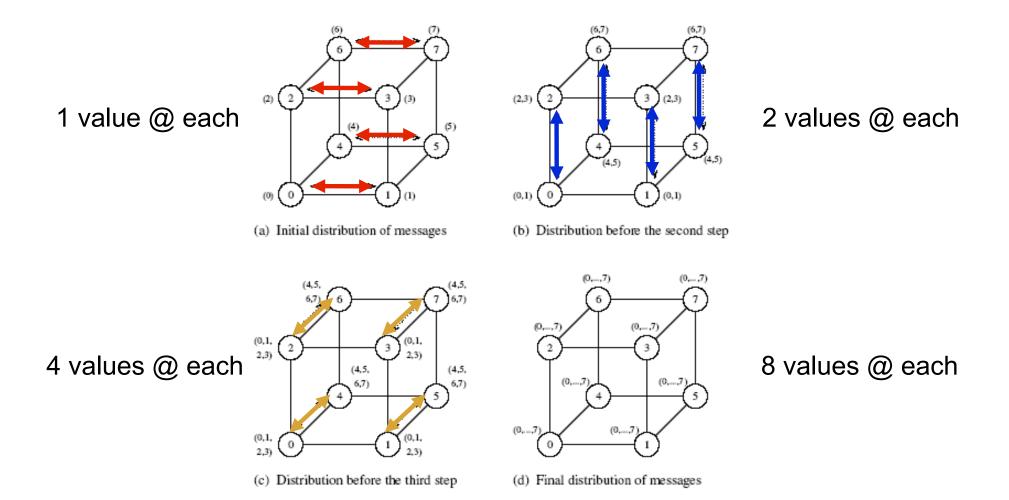
—each node collects  $p^{1/2}$  messages for nodes in its own row —consolidates into a single message of size  $mp^{1/2}$ 

• Perform column-wise all-to-all broadcast of merged messages



### **All-to-all Broadcast on a Hypercube**

- Generalization of the mesh algorithm to *log p* dimensions
- Message size doubles in each of *log p* steps



### **All-to-all Broadcast on a Hypercube**

1. **procedure** ALL\_TO\_ALL\_BC\_HCUBE(*my\_id*, *my\_msg*, *d*, *result*) 2. begin 3.  $result := my\_msg;$ 4. for i := 0 to d - 1 do 5. partner :=  $my_i d \text{ XOR } 2^i$ ; 6. send result to partner; 7. **receive** msg from partner; 8. result := result  $\cup$  msg; 9. endfor; 10. end ALL\_TO\_ALL\_BC\_HCUBE

#### **All-to-all Reduction**

- Similar to all-to-all broadcast, except for the merge
- Algorithm sketch

my\_result = local\_value

for each round send my\_result to partner receive msg my\_result = my\_result ⊕ msg

post condition: each my\_result now contains global result

#### **Cost Analysis for All-to-All Broadcast**

• Ring

 $-(t_s + t_w m)(p-1)$ 

- Mesh
  - $\begin{array}{ll} --phase \ 1: (t_s + t_w m)(p^{1/2} 1) \\ --phase \ 2: (t_s + t_w mp^{1/2})(p^{1/2} 1) \\ --total: \ 2t_s(p^{1/2} 1) + t_w m(p 1) \end{array}$
- Hypercube

$$T = \sum_{i=1}^{\log p} (t_s + 2^{i-1} t_w m)$$

$$= t_s \log p + t_w m(p-1).$$

#### Above algorithms are asymptotically optimal in msg size

#### **Prefix Sum**

- Pre-condition
  - —given *p* numbers  $n_0, n_1, \dots, n_{p-1}$  (one on each node)
    - node labeled k contains  $n_k$
- Problem statement

—compute the sums  $s_k = \sum_{i=0}^{k} n_i$  for all k between 0 and p-1

- Post-condition
  - node labeled k contains  $s_k$

### **Prefix Sum**

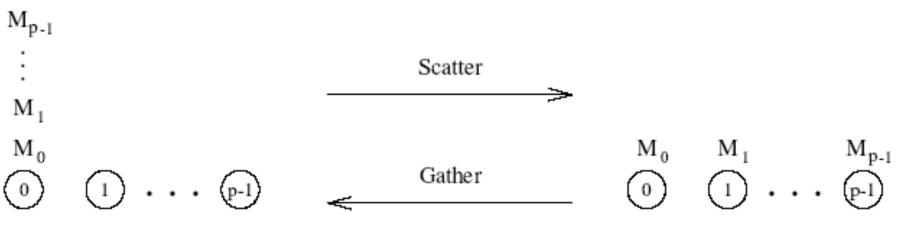
- Can use all-to-all reduction kernel to implement prefix sum
- Constraint
  - —prefix sums on node *k*: values from *k*-node subset with labels ≤ *k*
- Strategy
  - —implemented using an additional result buffer
  - —add incoming value to result buffer on node *k* 
    - only if the msg from a node  $\leq k$

## **Prefix Sum on a Hypercube**

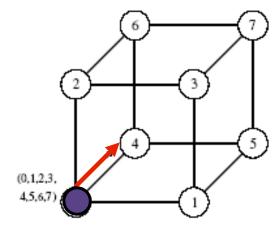
1.	<pre>procedure PREFIX_SUMS_HCUBE(my_id, my_number, d, resulf)</pre>
2.	begin
3.	result := my_number;
4.	msg := result;
5.	for $i := 0$ to $d - 1$ do
6.	partner := $my\_id$ XOR $2^i$ ;
7.	<b>send</b> msg to partner;
8.	<b>receive</b> number from partner;
9.	msg := msg + number;
10.	if (partner $< my_id$ ) then result := result + number;
11.	endfor;
12.	end PREFIX_SUMS_HCUBE

#### **Scatter and Gather**

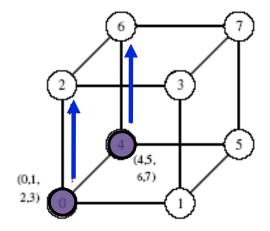
- Scatter
  - —a node sends a unique message of size *m* to every other node
    - AKA one-to-all personalized communication
  - -algorithmic structure is similar to broadcast
    - scatter: message size get smaller at each step
    - broadcast: message size stay constant
- Gather
  - -single node collects a unique message from each node
  - —inverse of the scatter operation; can be executed as such



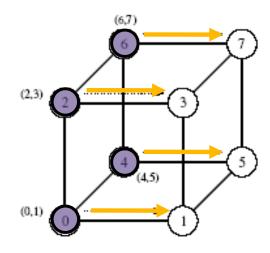
#### **Scatter on a Hypercube**



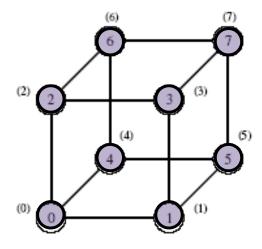
(a) Initial distribution of messages



(b) Distribution before the second step



(c) Distribution before the third step



(d) Final distribution of messages

#### **Cost of Scatter and Gather**

- Log p steps
  - —in each step
    - machine size halves
    - message size halves
- Time

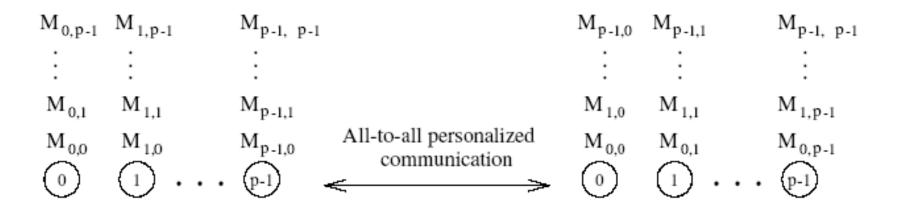
$$T = t_s \log p + t_w m(p-1).$$

• Note: time is asymptotically optimal in message size

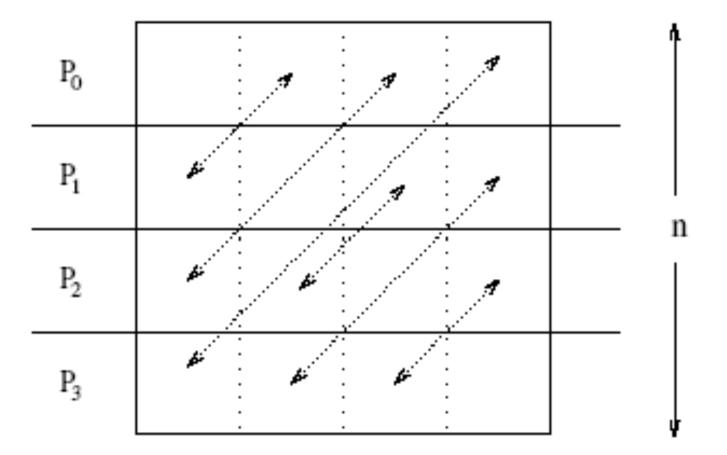
#### **All-to-All Personalized Communication**

#### **Total exchange**

• Each node: distinct message of size *m* for every other node



#### **All-to-All Personalized Communication**



#### **All-to-All Personalized Communication**

- Every node has p pieces of data, each of size m
- Algorithm sketch for a ring

for k = 1 to p - 1
send message of size m(p - k) to neighbor
select piece of size m out of message for self

Cost analysis

$$T = \sum_{i=1}^{p-1} (t_s + t_w m(p-i))$$
  
=  $t_s(p-1) + t_w m \sum_{i=1}^{p-1} i$   
=  $(t_s + t_w mp/2)(p-1)$ 

### **Optimizing Collective Patterns**

#### Example: one-to-all broadcast of large messages on a hypercube

- Consider broadcast of message *M* of size *m*, where *m* is large
- **Cost of straightforward strategy**  $T = (t_s + t_w m) \log p$
- Optimized strategy
  - —split *M* into *p* parts  $M_0$ ,  $M_1$ , ...  $M_p$  of size *m/p* each
    - want to place  $M_0 \cup M_1 \cup ... \cup M_p$  on all nodes
  - -scatter  $M_i$  to node i
  - -have nodes collectively perform all-to-all broadcast
    - each node k broadcasts its  $M_k$
- Cost analysis
  - -scatter time =  $t_s \log p + t_w (m/p)(p-1)$  (slide 27)
  - -all-to-all broadcast time =  $t_s \log p + t_w (m/p)(p-1)$  (slide 21)
  - -total time =  $2(t_s \log p + t_w(m/p)(p-1)) \approx 2(t_s \log p + t_w m)$ (faster than slide 13)

#### References

- Adapted from slides "Principles of Parallel Algorithm Design" by Ananth Grama
- Based on Chapter 4 of "Introduction to Parallel Computing" by Ananth Grama, Anshul Gupta, George Karypis, and Vipin Kumar. Addison Wesley, 2003