



# Design of Parallel Algorithms

Bulk Synchronous Parallel  
A Bridging Model of Parallel Computation

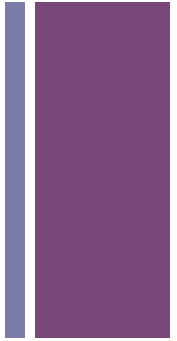
# + Need for a Bridging Model



- The RAM model has been reasonably successful for serial programming
  - The model provides a framework for describing the implementation of serial algorithms
  - The model provides reasonably accurate predictions for algorithm running times
- A bridging model is a model that can be used to design algorithms and also make reliable performance predictions
- Historically, there has not been a satisfactory bridging model for parallel computations. Either the model is good at describing algorithms (PRAM) or is good at describing performance (network model) but not both.
- Leslie Valiant proposed the BSP model as a potential bridging model
  - Basically an improvement on the PRAM model to incorporate more practical aspects of parallel hardware costs



# What is the Bulk Synchronous Parallel (BSP) model?



- Processors are coupled to local memories
- Communications happen in synchronized bulk operations
  - Data updates for the communications are inconsistent until the completion of a synchronization step
  - All of the communications that occur at the synchronization step are modeled in aggregate rather than tracking individual message transit times
- For data exchange, a one-sided communication model is advocated
  - E.g. data transfer through **put** or **get** operations that are executed by only one side of the exchange (as opposed to 2 sided where send-receive pairs must be matched up.)
- Similar to a coarse grained PRAM model, but exposes more realistic communication costs
- BSP provides realistic performance predictions



# Bulk Synchronous Parallel Programming



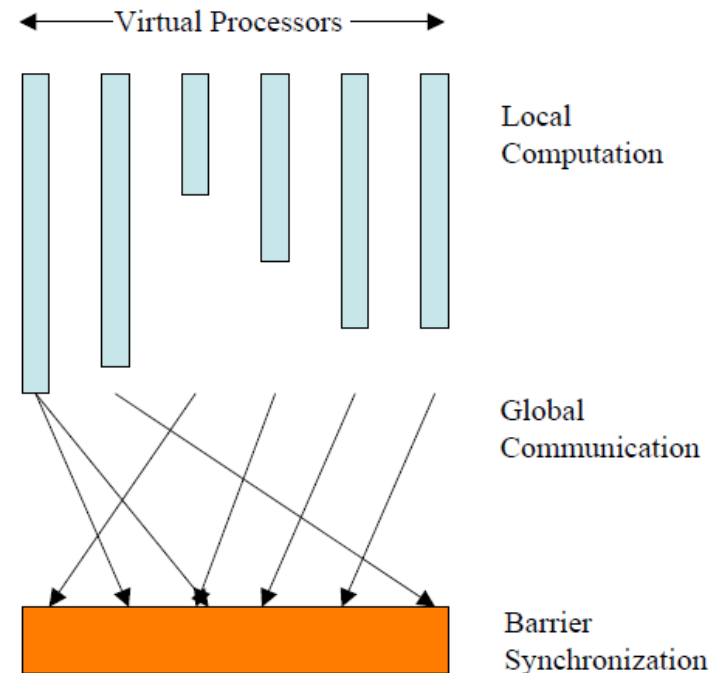
- Parallel Programs are developed through a series of super-steps
- Each super-step contains:
  - Computations that utilize local processor memory only
  - A communication pattern between processors called an h-relation
  - A barrier step whereby all (or subsets) of processors are synchronized
    - The communication pattern is not fully realized until the barrier step is complete
- The h-relation:
  - This describes communication pattern according to a single characteristic of the communication identified by the parameter called  $h$
  - $h$  is defined as the larger of the number of incoming or outgoing interactions that occur during the communication step
  - Time for communication is assumed to be  $mgh+l$  where  $m$  is the message size,  $g$  is an empirically determined bulk bandwidth factor, and  $l$  is an empirically determined time for barrier synchronization



# Architecture of a BSP Super-Step



- The super-step begins with local computations
- In some models, virtual processors are used to give the run-time system flexibility to balance load and communication
- Local computations are followed by a global communication step
- The global communications are completed with a barrier synchronization
- Since every super-step starts after the barrier, computations are time synchronized at the beginning of each super-step



# + Cost Model for BSP



- The network is defined by two bulk parameters
  - The parameter  $g$  represents the average per-processor rate of word transmission through the network. It is an analog to  $t_w$  in network models.
  - The parameter  $l$  is the time required to complete the barrier synchronization and represents the bulk latency of the network. It is an analog to  $t_s$  in network models.
- The cost of a super-step can be computed using the following formula
  - $t_{step} = \max(w_i) + mg \max(h_i) + l$
  - $w_i$  is the time for local work on processor  $i$
  - $h_i$  is the number of incoming or outgoing messages for processor  $i$
  - $m$  is the message size
  - $g$  is the machine specific BSP bandwidth parameter
  - $l$  is the machine specific BSP latency parameter



# Example of BSP implementations of broadcast (central scheme)



- Since there is no global shared memory in the BSP model, we need to broadcast a value before it can be used by all processors
- There are several ways to implement broadcast algorithms, a central scheme would perform the broadcast by using one super-step with one processor communicating with all other processors. This we call the central scheme.
- In this approach the  $h$  relation will be  $p-1$  since one processor will need to send a message to all other processors.
- The cost for this scheme is  $t_{central} = gh+l = g(p-1) + l$



# Example: BSP broadcast using binary tree scheme



- Broadcast using a tree approach where the algorithm proceeds in  $\log p$  steps
- Each step, every processor that presently has broadcast data sends to a processor that has no data
  - Processors that have broadcast data doubles in each step
- Since each processor either sends or receives one or no data each step, the  $h$  relation is always  $h=1$
- The time for each step of this algorithm is  $t_{step} = g+l$
- The time for the overall broadcast algorithm that includes all  $\log p$  steps
  - $t_{tree} = (g+l) \log p$





# Optimizing broadcasts under BSP



- The central algorithm time:
  - $t_{central} = g(p-1) + l$
- The tree algorithm time:
  - $t_{tree} = (g+l) \log p$
- If  $l \gg g$  then for sufficiently small  $p$ , then  $t_{central} < t_{tree}$
- Can we optimize broadcast for specific system where we know  $g$  and  $l$ ?
  - There is no reason that we are constrained only double in each step, We could triple, quadruple, or more each step.
  - Combining the central and tree algorithm can yield an algorithm that can be optimized for architecture parameters



# Cost of the hybrid broadcast algorithm



- Each step of the algorithm, processors that have data will communicate with  $k-1$  other processors, therefore  $h=k-1$  in each step
- After  $\log_k p$  steps, all processors will have shared the broadcast data
- Therefore the cost of each step of the hybrid algorithm is  $(k-1)g$  and so the cost of the hybrid algorithm is  $t_{hybrid} = ((k-1)g + l)\log_k p$
- To optimize set  $k$  such that  $t_{hybrid}'(k)=0$ , from this we find optimal  $k$  set by
  - $l/g = 1+k*(\ln(k)-1)$
- For a general message of  $m$  words, the broadcast algorithm can be shown to be  $t_{hybrid} = (m(k-1)g + l)\log_k p$ , and the optimal setting for  $k$  becomes
  - $l/(mg) = 1+k*(\ln(k)-1)$

# + Practical application of BSP



- Several parallel programming environments have been developed based on the BSP model
- The second generation of the MPI standard, MPI-2, has an extended its API to include a one-sided communication structure that can emulate the BSP model (e.g. it is one-sided + barrier synchronization)
- Even when using two sided communications, parallel programs are often developed as a sequence of super-steps. Using the BSP model, these can be analyzed using a bulk view of communications.
- The BSP model assumes that network is homogenous, but architectural changes, such as multi-core architectures, present challenges
  - Currently model is being extended to support hierarchical computing structures

# + Discussion Topic

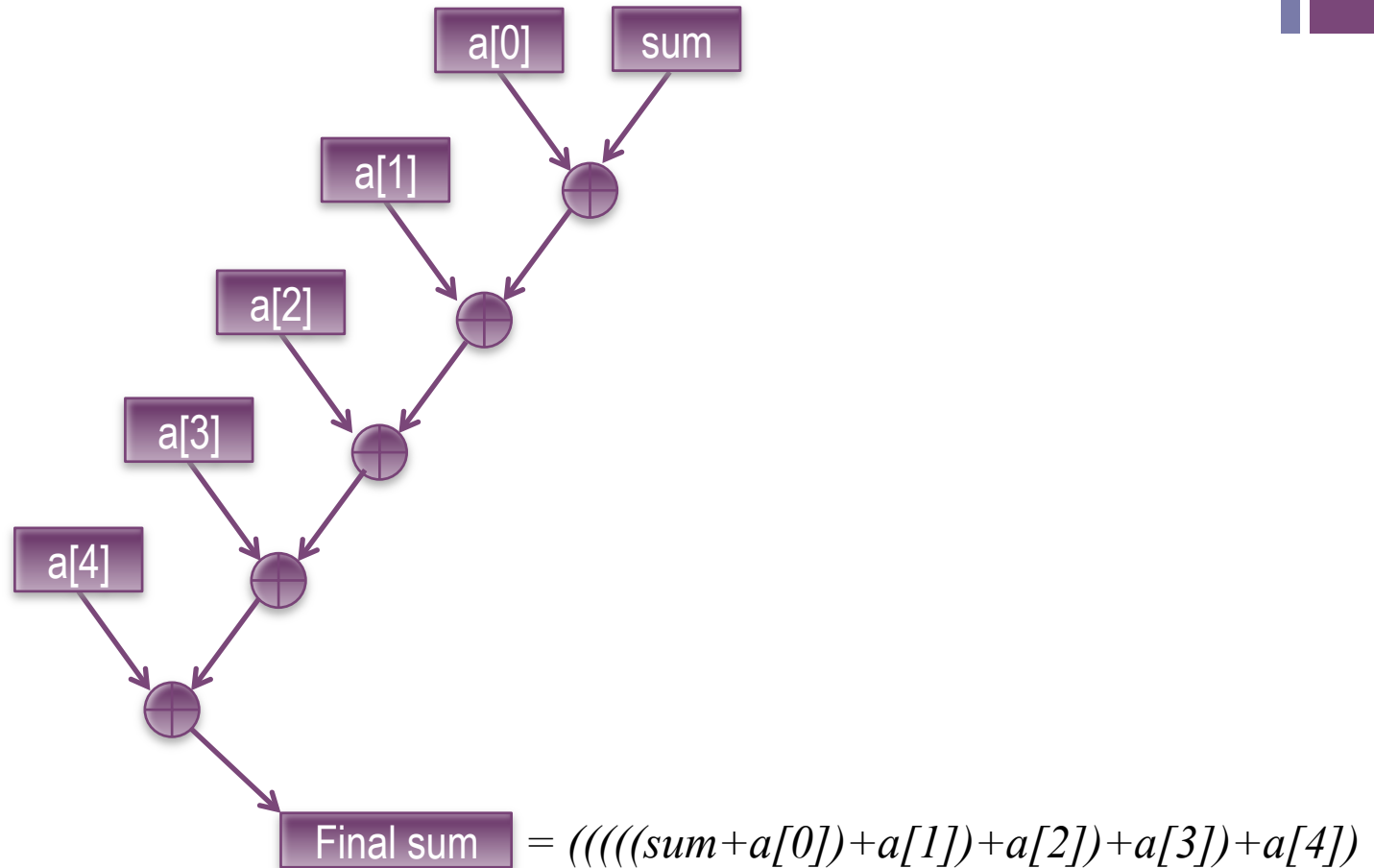


- Implementation of summing  $n$  numbers using BSP model
- Serial Implementation:

```
int sum = 0 ;  
for(int i=0;i<n;++i)  
    sum = sum + a[i] ;
```

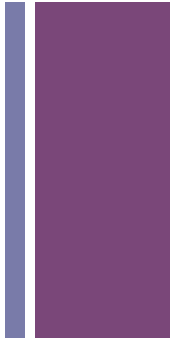
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# Dependency graph for serial summation





# Problems with parallelizing the serial code



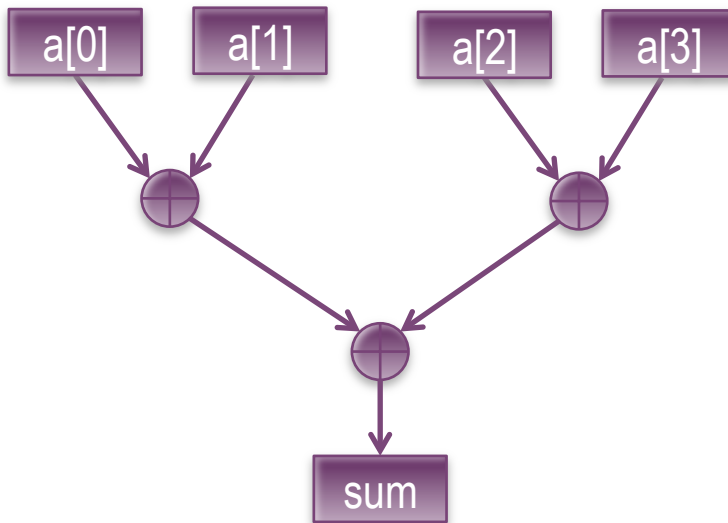
- The dependency graph does not allow one to perform subsequent operations.
  - It is not possible, as the algorithm is formulated, to execute additions in parallel
- We note that the addition operation is associative
  - NOTE! This is not true for floating point addition!
  - Although floating point addition is not associative, it is approximately associative
    - Accurately summing large numbers of floating point values, particularly in parallel, is a deep problem
    - For the moment we will assume floating point is associative as well, but note that in general an optimizing compiler cannot assume associativity of floating point operations!
- We can exploit associativity to increase parallelism



# How does associativity help with parallelization?



- We can recast the problem from a linear structure to a tree:
  - $((((a_0+a_1)+a_2)+a_3) = ((a_0+a_1)+(a_2+a_3))$
  - Now  $a_0+a_1$  and  $a_2+a_3$  can be performed concurrently!





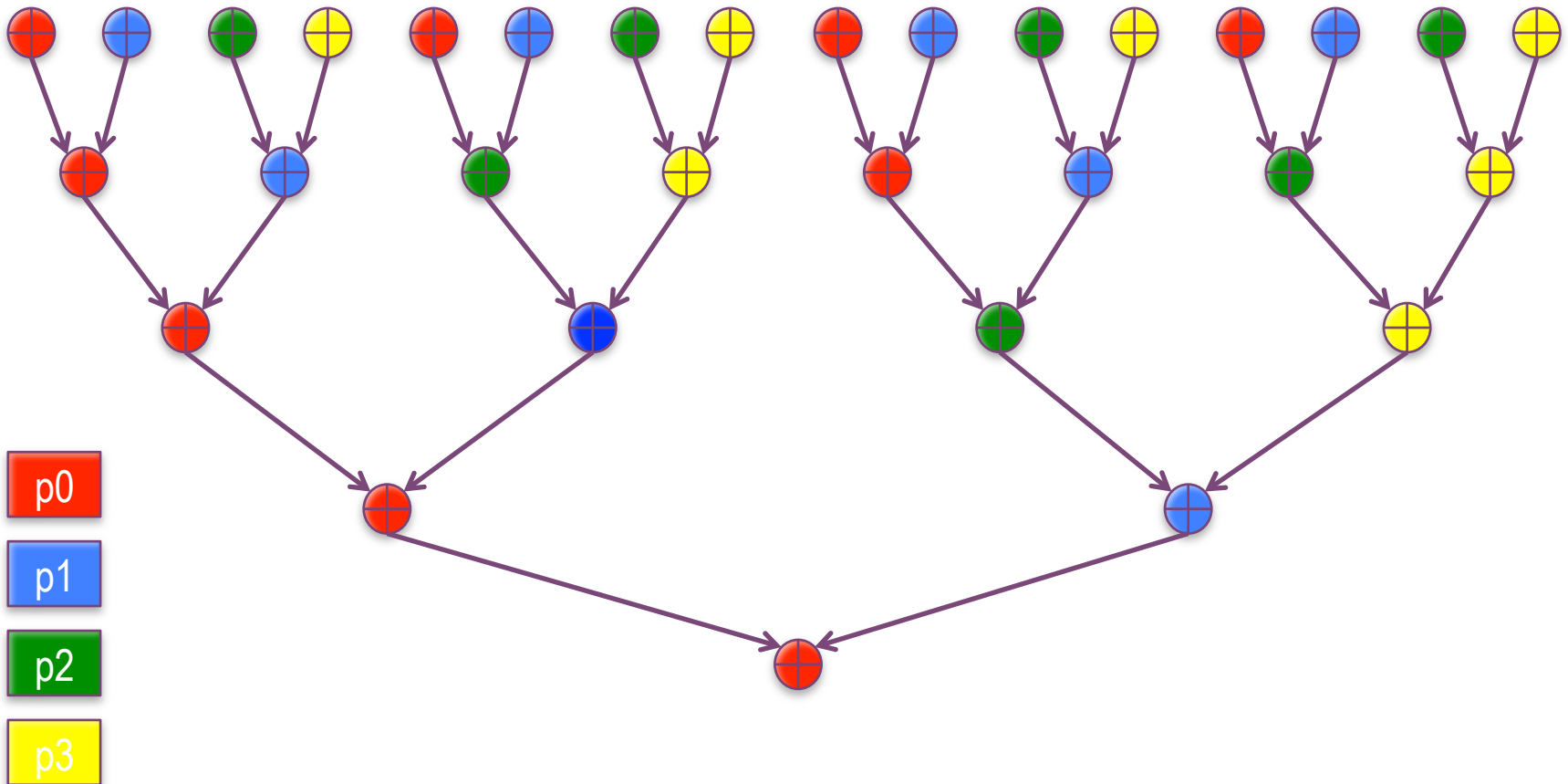
# What are the costs of this transformation



- Using operator associativity we are able to reveal additional parallelism, however there are costs
  - For the serial summing algorithm only one register is needed to store intermediate results (we used the sum variable)
  - For the tree based summing algorithm we will need to store  $n/2$  intermediate results for the first concurrent step
- For summing where  $2n \gg p$ , maximizing concurrency may introduce new problems:
  - Storing extra intermediate results increase memory requirements of algorithm and may overwhelm available registers
  - Assigning operations to processors (graph partitioning) is needed to parallelize the summation. Some mappings will introduce significantly more inter-processor communication than others



# + Mapping Operators to Processors Round Robin Allocation

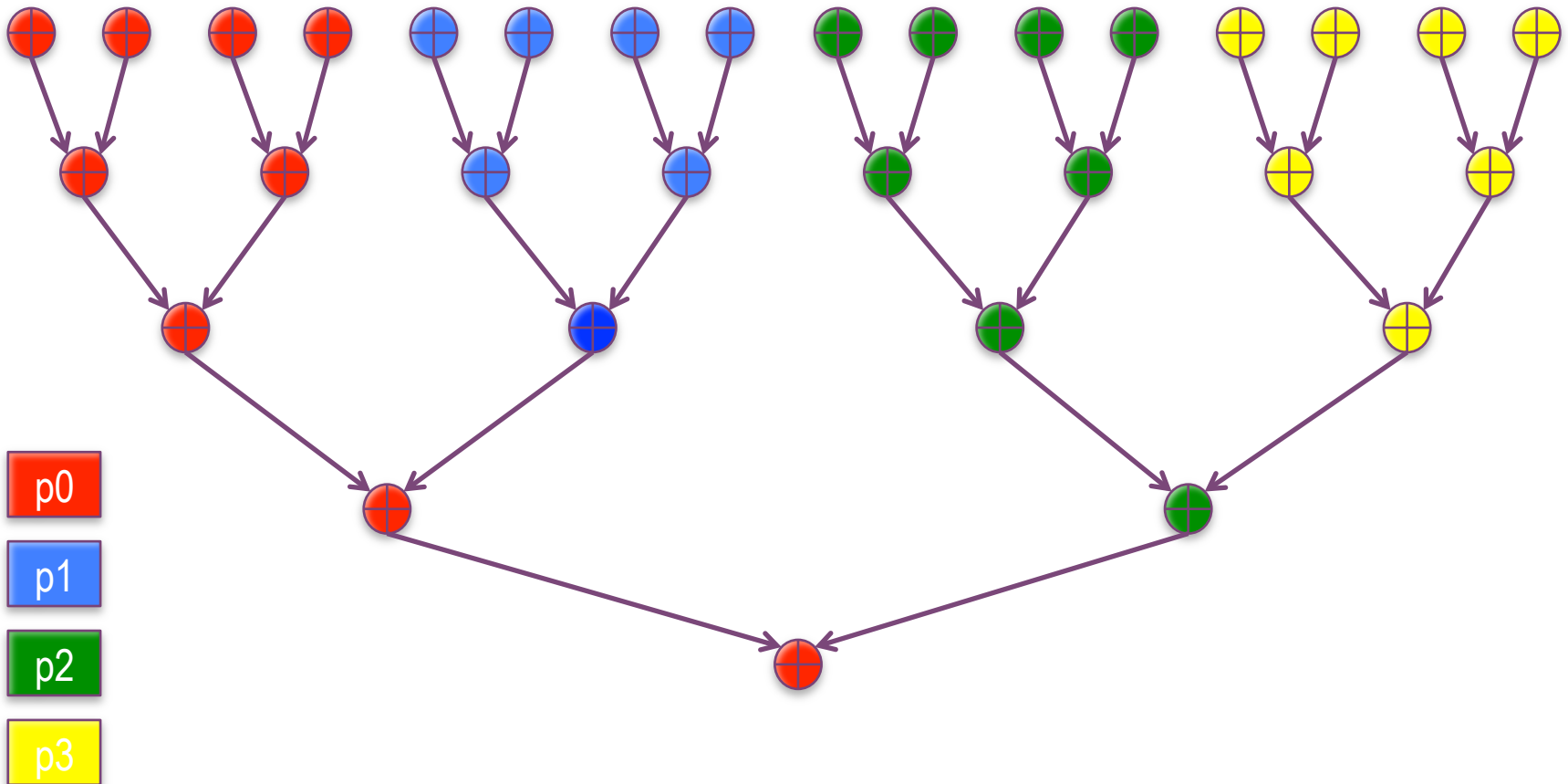


# + BSP model for round robin allocation of the tree

- Since there is communication for each level of the tree, there will be  $\log n$  super-steps in the algorithm
- For level  $i$  in the tree, the algorithm will perform  $\max(n/(2ip), 1)$  operations on at least one processor.
- For level  $i$  in the tree, the algorithm will utilize an  $h$  relation where  $h = \max(n/(2ip), 2)$
- Therefore the running time to sum  $n$  numbers on  $p$  processors using the BSP model is

$$t_{sum} = \sum_{i=1}^{\log n} \left\{ \left\lceil \frac{n}{2ip} \right\rceil t_c + \left\lceil \frac{n}{4ip} \right\rceil 2g + l \right\} \cong \frac{n}{p} (t_c + g) + l \log n$$

# + Mapping Operators to Processors Communication Minimizing Allocation





# BSP model for optimized allocation sum



- Notice that only the last  $\log p$  levels of the tree will require communication between processors, therefore there will be only  $\log p$  super-steps
- The first step will require  $n/p-1$  operations per processor, and the remaining steps will only require  $1$  operation
- During these final  $\log p$  steps, at most a processor either receives or send one piece of information, and so  $h = 1$  for the h-relation
- From this the BSP model running time can be derived:

$$t_{sum} = \left(\frac{n}{p} - 1\right)t_c + \sum_{i=1}^{\log p} \{t_c + g + l\} = \left(\frac{n}{p} - 1\right)t_c + (t_c + g + l)\log p$$

# + Comments on BSP analysis



- Obviously, in the BSP model, different allocations of work to processors can have radically different running times even though the work is equally balanced.
- For a PRAM model, both allocations would have had the same cost which is unrealistic.
- The cost structure of the BSP algorithms favors algorithms that have greater locality
- Even if we do not explicitly use a BSP model, we typically think of our algorithm going through a sequence of steps even if the implementation never explicitly enforces a barrier to get all processors to a unified state. Therefore the BSP model closely matches how we typically think about practical parallel programs.

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Q&A

