# Parallelism

**CS242** 

Lecture 16

### Concurrency vs. Parallelism

#### Concurrency

- Multiple threads of control
- May interleave at any granularity
  - Above atomic operations
- Makes sense even on a single processor
- A structuring device to deal with unpredictable latencies

#### Parallelism

- Multiple threads of control that execute at the same time
- On multiple hardware devices
- Inherently about performance

### Amdahl's Law

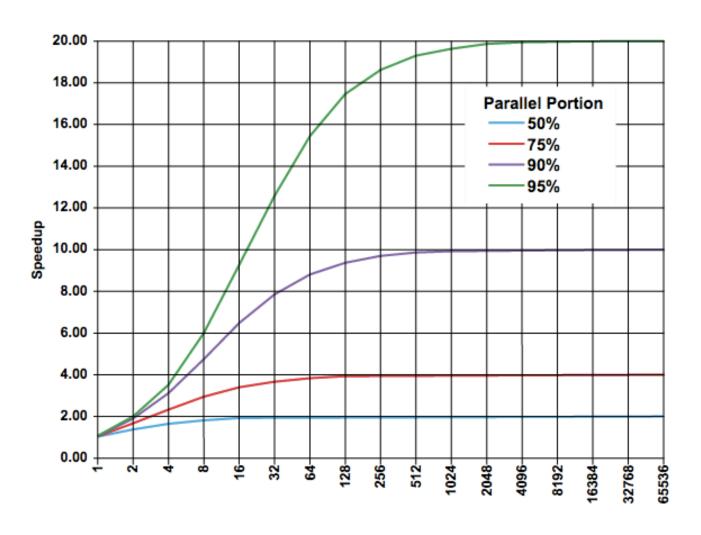
Speedup = 
$$\frac{1}{(1-p) + (p/s)}$$

#### where

p = portion of the program sped up

s = factor improvement of that portion

## Parallelism: Speed vs. Processors



### Discussion

- Amdahl's law is simple and general
  - Not about a specific machine or program
- And unforgiving
  - To speed up by 1000x, must parallelize 99.9%
  - To reach 10,000x, must parallelize 99.99%
  - And these are not very aggressive targets!

## Three Topics

• Bulk synchronous model

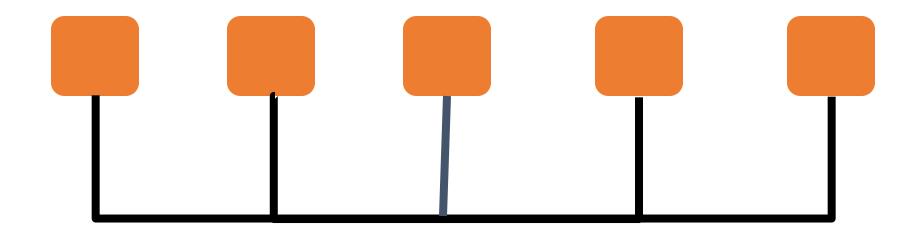
• SPMD

MapReduce

## Bulk Synchronous Model

- A model
  - An idealized machine
- Originally proposed for analyzing parallel algorithms
  - Leslie Valiant
  - "A Bridging Model for Parallel Computation", 1990

## The Machine



What are some properties of this machine model?

### Computations

• A sequence of *supersteps:* 

- Repeat:
- All processors do local computation
- Barrier all threads must reach the barrier before any proceed
- All processors communicate
- Barrier

What are properties of this computational model?

### Basic Properties

- Uniform
  - compute nodes
  - communication costs
- Separate communication and computation
- Synchronization is global

### The Idea

- Programs are
  - written for v virtual processors
  - run on p physical processors
- If  $v \ge p \log p$  then
  - Managing memory, communication and synchronization can be done automatically within a constant factor of optimal

### How Does It Work?

#### Roughly

- Memory addresses are hashed to a random location in the machine
- Guarantees that on average, memory accesses have the same cost
- The extra log p factor of threads are multiplexed onto the p processors to hide the latency of memory requests
- The processors are kept busy and do no more compute than necessary

# SPMD

## Terminology

- SIMD
  - Single Instruction, Multiple Data
- SPMD
  - Single Program, Multiple Data

## SIMD = Vector Processing

```
if (factor == 0)
  factor = 1.0
A[1..N] = B[1..N] * factor;
j += factor;
```

### Picture

#### Comments

- Single thread of control
  - Global synchronization at each program instruction
- Can exploit fine-grain parallelism
  - Assumption of hardware support

### SPMD = Single Program, Multiple Data

```
SIMD

if (factor == 0)
    factor = 1.0

A[1..N] = B[1..N] * factor;
    j += factor;

j += factor;
if (factor == 0)
    factor = 1.0

A[myid] = B[myid] * factor;

j += factor;
```

. . .

. . .

### Picture



#### Comments

- Multiple threads of control
  - One (or more) per processor
- Asynchronous
  - All synchronization is programmer-specified
- Threads are distinguished by myid
- Choice: Are variables local or global?

### Comparison

#### • SIMD

- Designed for tightly-coupled, synchronous hardware
- i.e., vector units

#### SPMD

- Designed for clusters
- Too expensive to synchronize every statement
- Need a model that allows asynchrony

#### MPI

- Message Passing Interface
  - A widely used standard
  - Runs on everything
- Some similarities to the Pi Calculus
  - For one-to-one communication
  - Also has one-to-all and all-to-all communication primitives

Most popular way to write SPMD programs

### MPI Programs

- Standard sequential programs
  - All variables are local to a thread

- Augmented with calls to the MPI interface
  - SPMD model
  - Every thread has a unique identifier
  - Threads can send/receive messages
  - Synchronization primitives

#### MPI Point-to-Point Routines

- MPI\_Send(buffer, count, type, dest, ...)
- MPI\_Recv(buffer, count, type, source, ...)

## Example

```
for (....) {
// p = number of chunks of 1D grid, id = process id, h[] = local chunk of the grid
// boundary elements of h[] are copies of neighbors boundary elements
 .... Local computation ...
// exchange with neighbors on a 1-D grid
if (0 < id)
   MPI_Send ( &h[1], 1, MPI_DOUBLE, id-1, 1, MPI_COMM_WORLD );
if ( id < p-1 )
   MPI_Recv ( &h[n+1], 1, MPI_DOUBLE, id+1, 1, MPI_COMM_WORLD, &status );
if ( id < p-1 )
   MPI_Send ( &h[n], 1, MPI_DOUBLE, id+1, 2, MPI_COMM_WORLD );
if ( 0 < id )
   MPI_Recv ( &h[0], 1, MPI_DOUBLE, id-1, 2, MPI_COMM_WORLD, &status );
... More local computation ...
```

### MPI Collective Communication Routines

- MPI\_Barrier(...)
- MPI\_Bcast(...)
- MPI\_Scatter(...)
- MPI\_Gather(...)
- MPI\_Reduce(...)

## Typical Structure

What does this remind you of?

## Bulk Synchronous/SPMD Model

Easy to understand

- Phase structure guarantees no data races
  - Barrier synchronization also easy to understand
- Fits many problems well

### **PGAS Model**

- PGAS = Partitioned Global Address Space
- There is one global address space
- But each thread owns a partition of the address space that is more efficient to access
  - i.e., the local memory of a processor
- Equivalent in functionality to MPI
  - But typically presented as a programming language
  - Examples: Split-C, UPC, Titanium

### PGAS Languages

No library calls for communication

• Instead, variables can name memory locations on other machines

```
// Assume y points to a remote location
// The following is equivalent to a send/receive
x = *y
```

### PGAS Languages

Also provide collective communication

Barrier

- Broadcast/Reduce
  - 1-many
- Exchange
  - All-to-all

### PGAS vs. MPI

- Programming model very similar
  - Both provide SPMD
- From a pragmatic point of view, MPI rules
  - Easy to add MPI to an existing sequential language
- For productivity, PGAS is better
  - Programs filled with low-level details of MPI calls
  - PGAS programs easier to modify
  - PGAS compilers can know more/do a better job

### Summary

- SPMD is well-matched to cluster programming
  - Also works well on shared memory machines
- One thread per core
  - No need for compiler to discover parallelism
  - No danger of overwhelming # of threads
- Model exposes memory architecture
  - Local vs. Global variables
  - Local computation vs. sends/receives

# MapReduce

### The Motivation

- Organizations today produce enormous amounts of data
  - Customer orders
  - Patient records
  - Social media interactions
  - Log files from all the computers
  - ...
- The only way to process this data is in parallel
  - Would take far too long on a single machine
  - Memory is the main the issue, not processing power

## The Programming Model

Programs have a type List(String) → a

- A string here is (substantial) block of data
  - Simply interpreted as a sequence of bytes; i.e., a string
  - Example: a file consisting of tens of megabytes to terabytes
- A program has the form

```
(reduce g) \diamond (map f) where f: String \rightarrow a and g: (a,a) \rightarrow a
```

#### Semantics

(reduce g) ◊ (map f)

#### What does • do?

- Group the pairs returned by the map on the first components
  - the *keys*
- Apply the reduction function to the set of second components for each key
  - The values
- Return the set of key-value pairs after the reduction

### Example: Word Count

Count the number of occurrences of each word in a corpus.

```
(reduce +) ♦ map ((map λw.(w,1)) o split '')
```

where split: Char  $\rightarrow$  String  $\rightarrow$  List(String)

#### Discussion

- A combinator-based language!
  - Recall SKI
- The combinator is perhaps unexpected
  - More than function composition
- Makes it easy to write parallel, distributed computations
  - Without knowing anything about parallelism, different kinds of parallel hardware, synchronization, etc.
  - Just need to understand basic functional programming

### Discussion

- Why is MapReduce functional?
- Using state makes the most sense when there is a hardware-provided global state
  - i.e., a single address space
- But in distributed machines, there is no single address space
  - Or rather, it is expensive to provide one in software
- Functional programming makes sense regardless of the scale of the underlying machine

### Evolution

- MapReduce became very popular
  - Gave programmers an easy way to write simple computations that they otherwise couldn't write themselves

- Downside: Many algorithms can't be expressed in MapReduce
  - At least efficiently
- Spark is a well-known successor to MapReduce

### Summary

- Two very different parallel programming models
- Bulk synchronous/SPMD
  - Message-passing based
  - Run many copies of the same program in parallel
  - Threads are distinguished by a myid unique to each thread
  - Threads communicate via message passing
  - Used mostly for scientific programming

#### MapReduce

- Based on functional combinators with obvious parallel implementations
- More expressive than simple composition because of the flexibility of the group-by operation that MapReduce adds between the map and reduce stages
- Used for data analytics