

CS 596: Introduction to Parallel Computing

Topic: Parallel Computing Architectures

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Posted: 01/30/17
Updated: 01/30/17

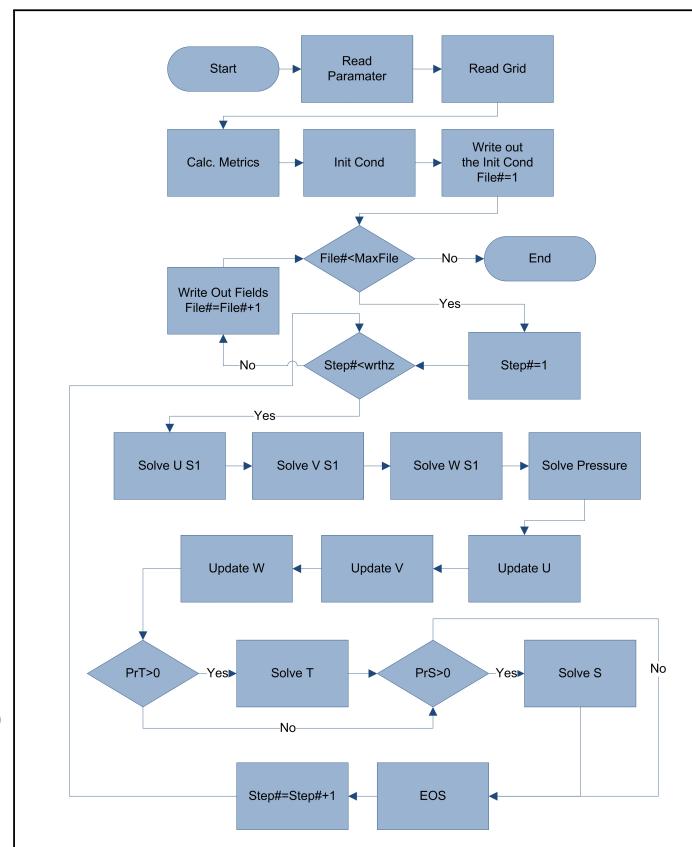
My serial code is working, why should I parallelize it?

- Compute bound
 - Large number of iterations
 - Long time to converge
- Data bound
 - Memory footprint is larger than RAM
(common for science apps)
 - Output may take FOREVER to
save/dump/transfer
- contention for resources/other processes
 - What other processes are running on the
machine?
 - What other data is running on the network?
 - Timeout problems?



General Curvilinear Environmental Model

- Simulate processes along the coast
- Serial version, Simple Seamount, appx 8k lines
 - 1 km /cell, 100x50x50 (small job)
 - large memory: 100 arrays $\tilde{O}(10^7)$
 - 200k iters ≈ 6 hours simulation time
 - **T_{wall} ≈ 70 hours $\approx O(10^{14})$ ops**
- Typical simulations need to run for days/weeks/months/years
 - Jaguar (ORNL), 2×10^5 cores (≈ 3 GHz, 16GB), 1.75×10^{12} flops
 - Could run this job in 0.3×10^{-2} seconds (note: est. does not include time to fire up 200k cores)
 - **Motivation for parallel computing!**



Source: <http://www.xsede.org>

Porting Code to Parallel has Many Challenges

- I have 200k cores, now what do I do with them?
- How to distribute the data? Is it memory/IO bound)?
- How to distribute the computation? Is it compute bound?
- How do you synchronize the results?

What is Parallel Computing and Why do We Need it?

- Large problems:
 - Biz: BM has 433,000 employees; how do you update paycheck/employee data?
 - Science: Hurricane Predictions, GoM is appx 800×800 km = 6.4×10^5 points/array. Typically have 10^2 arrays. Forecast models use multiple models, multiple clusters, take 12 hours (wall clock) to forecast out 3-4 days on parallel computers (probably 500 nodes). See <http://www.srh.noaa.gov/ssd/nwpmodel/html/nhcmodel.htm>, f
- Large data: These models generate massive data sets (Tera-to-peta-bytes per simulation).
 - My coastal ocean model generates Vel(U,V,W), Pressure,Temp,Salinity files once each iteration at 2.5MB/file. 6 hour test run is 2×10^5 iterations = 3×10^6 MBytes (but we don't have room so we only store 100 slices).
 - Where do you store the data?
 - How do you move it? Sneaker Net?
- Fast networks are required
- Power consumption is big issue

Typical HPC Computing Resources

- Compute
 - HPC systems & Clusters
 - GPU
 - Clouds
- Archival & Storage
 - high speed servers with large storage
 - Clouds
- Network
- Visualization
- Cyberinfrastructure

HPC Units of Measure

What is the scale of modern parallel resources & computations

Unit	Definition
Flops	floating point operations
Flop/s	floating point operations per second

Prefix	#Flops/sec	Bytes	10^n	2^n
Mega	Mflop/s	Mbyte	10^6	$2^{20} = 1,048,576$
Giga	Gflop/s	Gbyte	10^9	$2^{30} = 1,073,741,824$
Tera	Tflop/s	Tbyte	10^{12}	$2^{40} = 10,995,211,627,776$
Peta	Pflop/s	Pbyte	10^{15}	$2^{50} = 1,125,899,906,842,624$
Exa	Eflop/s	Ebyte	10^{18}	$2^{60} = 1.15292152^{18}$

A Large HPC Machine: Kraken

Hostname	kraken.nics.teragrid.org
Manufacturer	Cray
Model	XT5
Processor Cores	112896
Nodes	9408
Memory	147.00 TB
Peak Performance	1174.00 TFlops
Disk	2400.00 TB



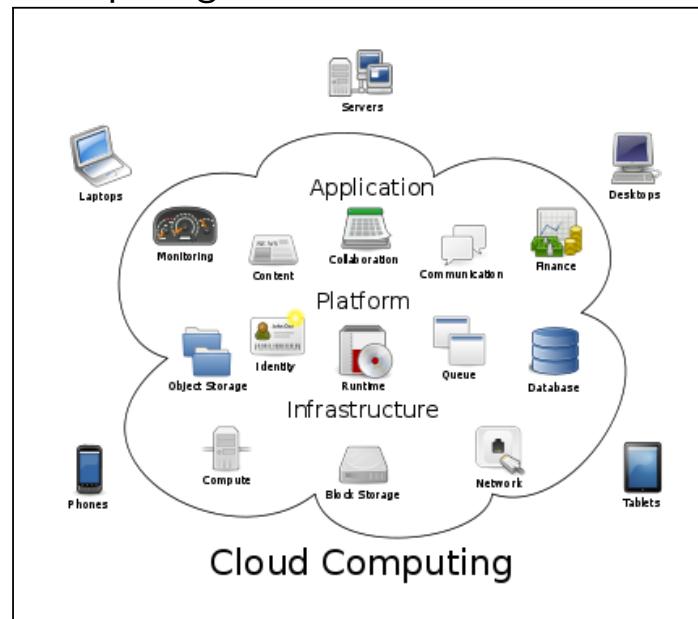
Source: <http://www.xsede.org>

Cloud Computing

Cloud computing is the use of computing resources (hardware and software) that are delivered as a service over a network (typically the Internet).

There are many types of public cloud computing:

- Infrastructure as a service (IaaS),
- Platform as a service (PaaS),
- Software as a service (SaaS)
- Storage as a service (STaaS)
- Security as a service (SECaaS)
- Data as a service (DaaS)
- Business process as a service (BPaaS)
- Test environment as a service (TEaaS)
- Desktop as a service (DaaS)
- API as a service (APIaaS)
- Cyberinfrastructure



Source: http://en.wikipedia.org/wiki/Cloud_computing

Cyberinfrastructure (CI)

- Cyberinfrastructure - Research environment or infrastructure based upon distributed computer, information and communication technology. (From: Atkins PITCAC Report, 2003)
- Resources: data acquisition, data storage, data management, data integration, data mining, data visualization, computing and information processing services distributed over the Internet beyond the scope of a single institution.*
- Scientific Comptuing: a technological and sociological solution to the problem of efficiently connecting laboratories, data, computers, and people with the goal of enabling derivation of novel scientific theories and knowledge.*
- Cyberinfrastructure = HPC + Cloud/Grid + Networks

* Source: <http://en.wikipedia.org/wiki/Cyberinfrastructure>

HPC Requirements

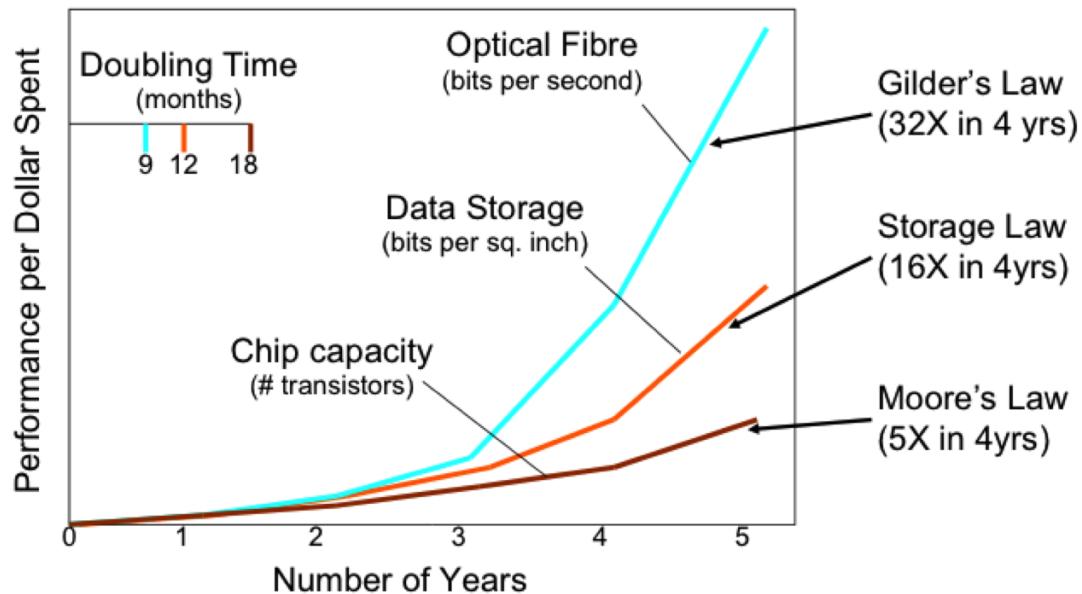
- Supercomputers: Moores Law - compute cycles exceeding ability to move data
 - Networks needed to move data around to resources
 - Large data sets (GB in 96, but now peta-bytes)
- Computational apps will always use up resources:
 - Just increase resolution: e.g. $3 \times 3 \times 3 = 27 - > 5 \times 5 \times 5 = 125$
- New architectures impacting HPC libraries, programming models, languages
- Security: desire to protect national resources
- NSF (and other gov agencies) spend money to build infrastructure

Technology Growth Laws

- **Moore's Law:** Processing power of a cpu/core doubles every 18 months; corollary, computers become faster and the price of a given level of computing power halves every 18 months (Gordon Moore, 70's).
- **Gilder's Law:** Total bandwidth of communication systems triples every twelve months (George Gilder).
- **Metcalfe's Law:** The value of a network is proportional to the square of the number of nodes; so, as a network grows, the value of being connected to it grows exponentially, while the cost per user remains the same or even reduces (Robert Metcalfe, originator of Ethernet)
- **Kryder's Law:** The density of information on hard drives growing faster than Moore's law, and is doubling roughly every 13 months.

Source: <http://www.jimpinto.com/writings/techlaws.html>, and
<http://www.scientificamerican.com/article/kryders-law/>

Technology Growth Laws



Source: G. Stix, Triumph of Light. Scientific American. January 2001

HPC Performance

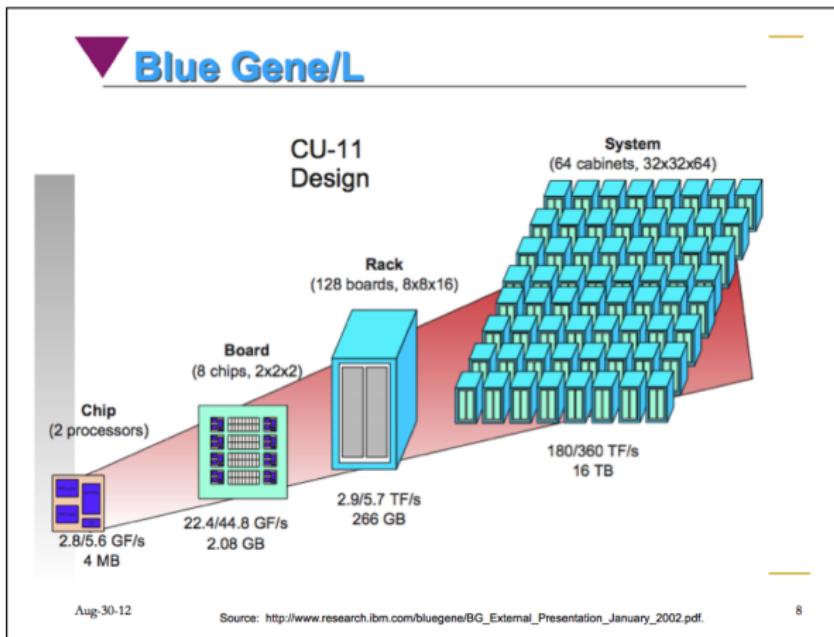


Source: <http://www.top500.org>

Top500



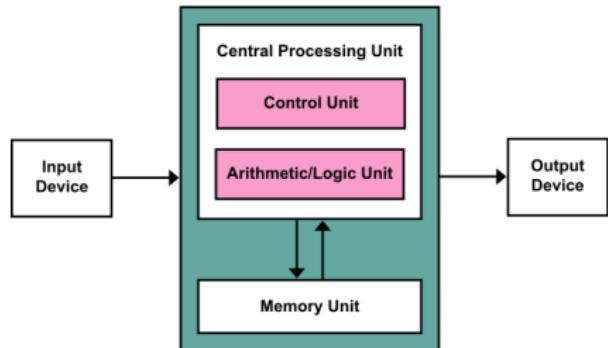
Figure: Source: <http://blog.infinibandta.org/tag/petascale/>



HPC Hardware: Blue Gene/L Hardware

Von Neumann electronic digital computer

- Central processing unit:
 - arithmetic logic unit (ALU)
 - processor registers
- Control unit:
 - instruction register
 - program counter
- Memory unit:
 - data
 - instructions
- External mass storage
- Input and output mechanisms



Source: http://en.wikipedia.org/wiki/Von_Neumann_architecture

The von Neumann Architecture

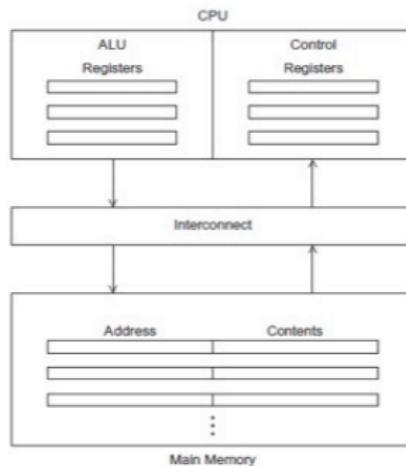
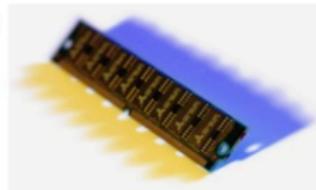


Figure 2.1

Main memory

- This is a collection of locations, each of which is capable of storing both instructions and data.
- Every location consists of an address, which is used to access the location, and the contents of the location.



Central processing unit (CPU)

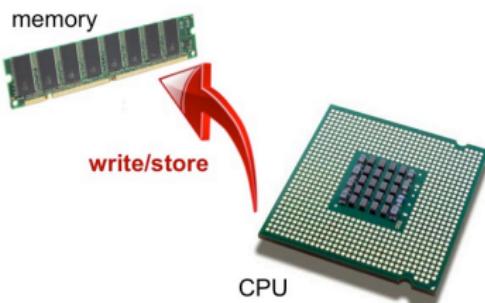
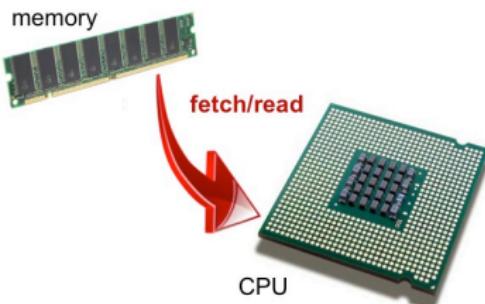
- Divided into two parts.
- **Control unit** - responsible for deciding which instruction in a program should be executed. (*the boss*)
- **Arithmetic and logic unit (ALU)** - responsible for executing the actual instructions. (*the worker*)



Key terms

- **Register** – very fast storage, part of the CPU.
- **Program counter** – stores address of the next instruction to be executed.
- **Bus** – wires and hardware that connects the CPU and memory.





An operating system “process”

- An instance of a computer program that is being executed.
- Components of a process:
 - The executable machine language program.
 - A block of memory.
 - Descriptors of resources the OS has allocated to the process.
 - Security information.
 - Information about the state of the process.

Multitasking

- Gives the illusion that a single processor system is running multiple programs simultaneously.
- Each process takes turns running. (**time slice**)
- After its time is up, it waits until it has a turn again. (**blocks**)

Threading

- Threads are contained within processes.
- They allow programmers to divide their programs into (more or less) independent tasks.
- The hope is that when one thread blocks because it is waiting on a resource, another will have work to do and can run.

Parallel Hardware Architectures

Computer Architecture Background

/

A process and two threads

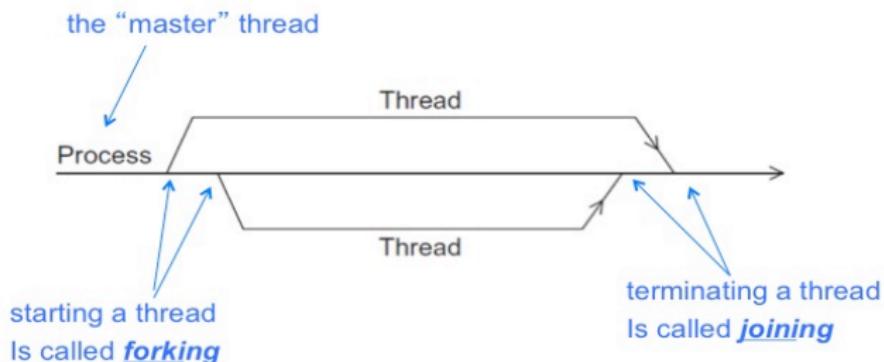
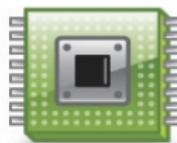


Figure 2.2



MODIFICATIONS TO THE VON NEUMANN MODEL

Basics of caching

- A collection of memory locations that can be accessed in less time than some other memory locations.
- A CPU cache is typically located on the same chip, or one that can be accessed much faster than ordinary memory.



Principle of locality

- Accessing one location is followed by an access of a nearby location.
- **Spatial locality** – accessing a nearby location.
- **Temporal locality** – accessing in the near future.

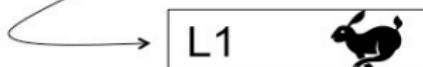
Principle of locality

```
float z[1000];  
...  
sum = 0.0;  
for (i = 0; i < 1000; i++)  
    sum += z[i];
```



Levels of Cache

smallest & fastest



L2

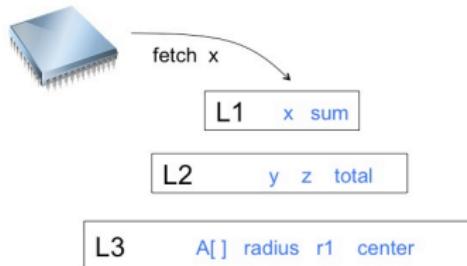


L3

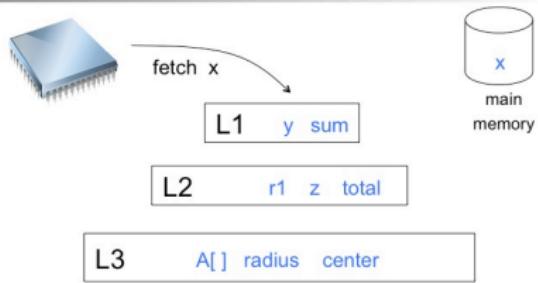


largest & slowest

Cache hit



Cache miss



Issues with cache

- When a CPU writes data to cache, the value in cache may be inconsistent with the value in main memory.
- **Write-through** caches handle this by updating the data in main memory at the time it is written to cache.
- **Write-back** caches mark data in the cache as **dirty**. When the cache line is replaced by a new cache line from memory, the **dirty** line is written to memory.

Cache mappings

- **Full associative** – a new line can be placed at any location in the cache.
- **Direct mapped** – each cache line has a unique location in the cache to which it will be assigned.
- **n -way set associative** – each cache line can be placed in one of n different locations in the cache.

n-way set associative

- When more than one line in memory can be mapped to several different locations in cache we also need to be able to decide which line should be replaced or **evicted**.



Example

Memory Index	Cache Location		
	Fully Assoc	Direct Mapped	2-way
0	0, 1, 2, or 3	0	0 or 1
1	0, 1, 2, or 3	1	2 or 3
2	0, 1, 2, or 3	2	0 or 1
3	0, 1, 2, or 3	3	2 or 3
4	0, 1, 2, or 3	0	0 or 1
5	0, 1, 2, or 3	1	2 or 3
6	0, 1, 2, or 3	2	0 or 1
7	0, 1, 2, or 3	3	2 or 3
8	0, 1, 2, or 3	0	0 or 1
9	0, 1, 2, or 3	1	2 or 3
10	0, 1, 2, or 3	2	0 or 1
11	0, 1, 2, or 3	3	2 or 3
12	0, 1, 2, or 3	0	0 or 1
13	0, 1, 2, or 3	1	2 or 3
14	0, 1, 2, or 3	2	0 or 1
15	0, 1, 2, or 3	3	2 or 3

Table 2.1: Assignments of a 16-line main memory to a 4-line cache

Caches and programs

```
double A[MAX][MAX], x[MAX], y[MAX];  
. . .  
/* Initialize A and x, assign y = 0 */  
. . .  
/* First pair of loops */  
for (i = 0; i < MAX; i++)  
    for (j = 0; j < MAX; j++)  
        y[i] += A[i][j]*x[j];  
. . .  
/* Assign y = 0 */  
. . .  
/* Second pair of loops */  
for (j = 0; j < MAX; j++)  
    for (i = 0; i < MAX; i++)  
        y[i] += A[i][j]*x[j];
```

Cache Line	Elements of A			
0	A[0][0]	A[0][1]	A[0][2]	A[0][3]
1	A[1][0]	A[1][1]	A[1][2]	A[1][3]
2	A[2][0]	A[2][1]	A[2][2]	A[2][3]
3	A[3][0]	A[3][1]	A[3][2]	A[3][3]



Virtual memory (1)

- If we run a very large program or a program that accesses very large data sets, all of the instructions and data may not fit into main memory.
- Virtual memory functions as a cache for secondary storage.

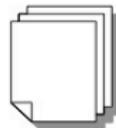
Virtual memory (2)

- It exploits the principle of spatial and temporal locality.
- It only keeps the active parts of running programs in main memory.

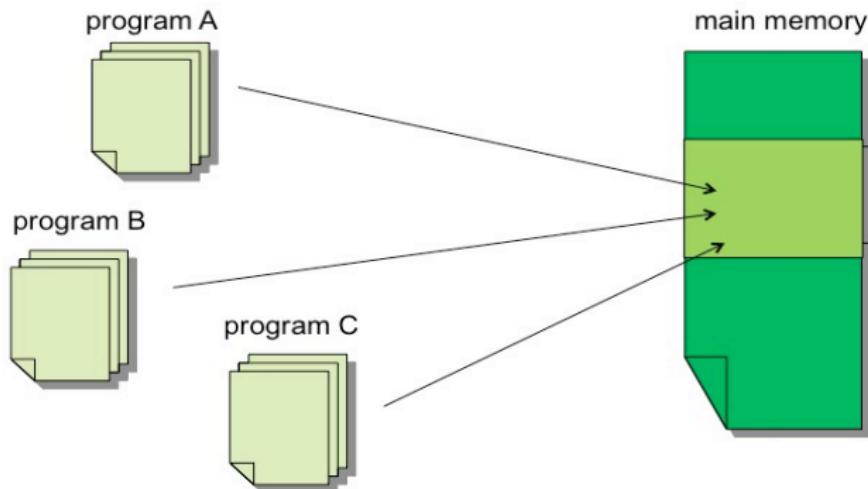
Virtual memory (3)

- **Swap space** - those parts that are idle are kept in a block of secondary storage.

- **Pages** – blocks of data and instructions.
 - Usually these are relatively large.
 - Most systems have a fixed page size that currently ranges from 4 to 16 kilobytes.



Virtual memory (4)



Virtual page numbers

- When a program is compiled its pages are assigned *virtual* page numbers.
- When the program is run, a table is created that maps the virtual page numbers to physical addresses.
- A **page table** is used to translate the virtual address into a physical address.

Page table

Virtual Address										
Virtual Page Number						Byte Offset				
31	30	...	13	12		11	10	...	1	0
1	0	...	1	1		0	0	...	1	1

Table 2.2: Virtual Address Divided into
Virtual Page Number and Byte Offset

Translation-lookaside buffer (TLB)

- Using a page table has the potential to significantly increase each program's overall run-time.
- A special address translation cache in the processor.

Translation-lookaside buffer (2)

- It caches a small number of entries (typically 16–512) from the page table in very fast memory.
- **Page fault** – attempting to access a valid physical address for a page in the page table but the page is only stored on disk.

Instruction Level Parallelism (2)

- **Pipelining** - functional units are arranged in stages.
- **Multiple issue** - multiple instructions can be simultaneously initiated.

Pipelining



Pipelining example (1)

Time	Operation	Operand 1	Operand 2	Result
1	Fetch operands	9.87×10^4	6.54×10^3	
2	Compare exponents	9.87×10^4	6.54×10^3	
3	Shift one operand	9.87×10^4	0.654×10^4	
4	Add	9.87×10^4	0.654×10^4	10.524×10^4
5	Normalize result	9.87×10^4	0.654×10^4	1.0524×10^5
6	Round result	9.87×10^4	0.654×10^4	1.05×10^5
7	Store result	9.87×10^4	0.654×10^4	1.05×10^5

Add the floating point numbers
 9.87×10^4 and 6.54×10^3

Pipelining example (2)

```
float x[1000], y[1000], z[1000];
.
.
.
for (i = 0; i < 1000; i++)
    z[i] = x[i] + y[i];
```

- Assume each operation takes one nanosecond (10^{-9} seconds).
- This for loop takes about 7000 nanoseconds.

Pipelining (3)

- Divide the floating point adder into 7 separate pieces of hardware or functional units.
- First unit fetches two operands, second unit compares exponents, etc.
- Output of one functional unit is input to the next.

Pipelining (4)

Time	Fetch	Compare	Shift	Add	Normalize	Round	Store
0	0						
1	1	0					
2	2	1	0				
3	3	2	1	0			
4	4	3	2	1	0		
5	5	4	3	2	1	0	
6	6	5	4	3	2	1	0
:	:	:	:	:	:	:	:
999	999	998	997	996	995	994	993
1000		999	998	997	996	995	994
1001			999	998	997	996	995
1002				999	998	997	996
1003					999	998	997
1004						999	998
1005							999

Table 2.3: Pipelined Addition.
Numbers in the table are subscripts of operands/results.

Pipelining (5)

- One floating point addition still takes 7 nanoseconds.
- But 1000 floating point additions now takes 1006 nanoseconds!



Multiple Issue (1)

- Multiple issue processors replicate functional units and try to simultaneously execute different instructions in a program.

```
for (i = 0; i < 1000; i++)
```

$$z[i] = x[i] + y[i];$$



Multiple Issue (2)

- **static** multiple issue - functional units are scheduled at compile time.
- **dynamic** multiple issue – functional units are scheduled at run-time.



superscalar

Speculation (1)

- In order to make use of multiple issue, the system must find instructions that can be executed simultaneously.
- In speculation, the compiler or the processor makes a guess about an instruction, and then executes the instruction on the basis of the guess.



Speculation (2)

$z = x + y ;$

$\text{if } (z > 0)$

$w = x ;$

else

$w = y ;$



If the system speculates incorrectly,
it must go back and recalculate $w = y$.

Hardware multithreading (1)

- There aren't always good opportunities for simultaneous execution of different threads.
- Hardware multithreading provides a means for systems to continue doing useful work when the task being currently executed has stalled.
 - Ex., the current task has to wait for data to be loaded from memory.

Hardware multithreading (2)

- **Fine-grained** - the processor switches between threads after each instruction, skipping threads that are stalled.
 - Pros: potential to avoid wasted machine time due to stalls.
 - Cons: a thread that's ready to execute a long sequence of instructions may have to wait to execute every instruction.

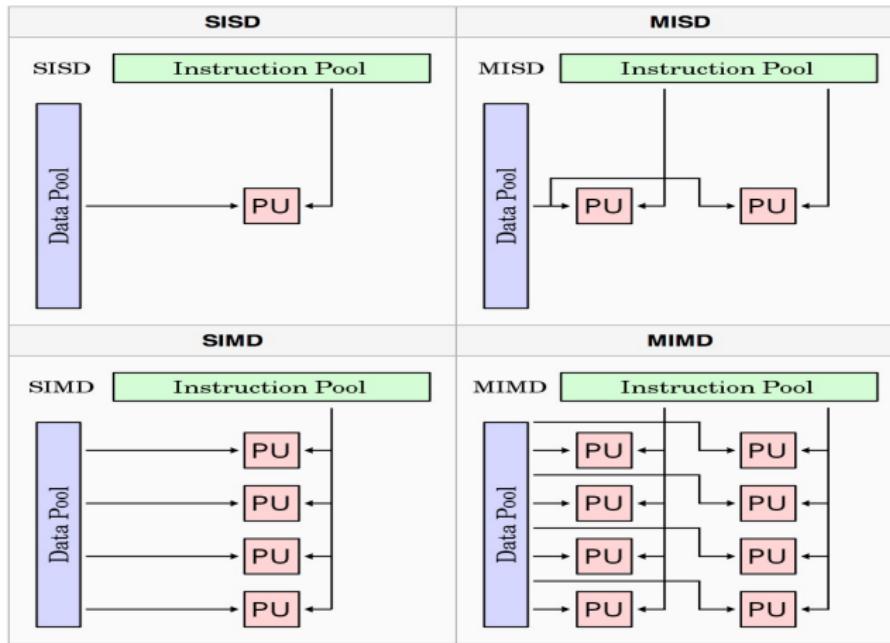
Hardware multithreading (3)

- **Coarse-grained** - only switches threads that are stalled waiting for a time-consuming operation to complete.

- Pros: switching threads doesn't need to be nearly instantaneous.
- Cons: the processor can be idled on shorter stalls, and thread switching will also cause delays.

Hardware multithreading (3)

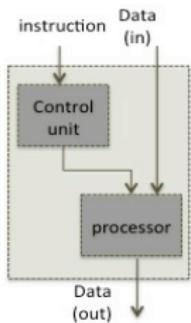
- **Simultaneous multithreading (SMT) - a variation on fine-grained multithreading.**
- Allows multiple threads to make use of the multiple functional units.



Source:

http://en.wikipedia.org/wiki/Flynn%27s_taxonomy

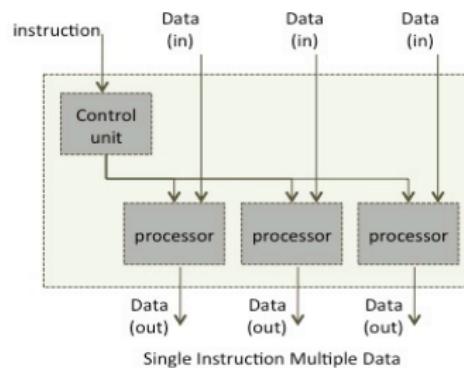
Single Instruction Single Data



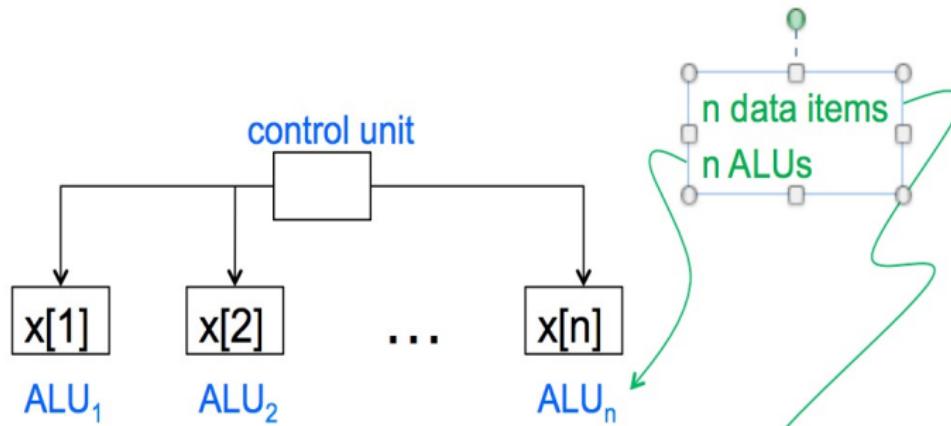
Single Instruction Single Data

Single Instruction Multiple Data

- Parallelism achieved by dividing data among the processors.
- Applies the same instruction to multiple data items.
- Called data parallelism



SIMD example



for ($i = 0; i < n; i++$)

$x[i] += y[i];$

SIMD Data Distribution

- What if we don't have as many ALUs as data items?
- Divide the work and process iteratively.
- Ex. $m = 4$ ALUs and $n = 14$ data items.

Round	ALU1	ALU2	ALU3	ALU4
1	X[0]	X[1]	X[2]	X[3]
2	X[4]	X[5]	X[6]	X[7]
3	X[8]	X[9]	X[10]	X[11]
4	X[12]	X[13]		

SIMD drawbacks

- All ALUs are required to execute the same instruction, or remain idle.
- In classic design, they must also operate synchronously.
- The ALUs have no instruction storage.
- Efficient for large data parallel problems, but not other types of more complex parallel problems.



Vector processors (1)

- Operate on arrays or vectors of data while conventional CPU's operate on individual data elements or scalars.
- Vector registers.
 - Capable of storing a vector of operands and operating simultaneously on their contents.



Vector processors (2)

- Vectorized and pipelined functional units.
 - The same operation is applied to each element in the vector (or pairs of elements).
- Vector instructions.
 - Operate on vectors rather than scalars.



Vector processors (3)

- Interleaved memory.
 - Multiple “banks” of memory, which can be accessed more or less independently.
 - Distribute elements of a vector across multiple banks, so reduce or eliminate delay in loading/storing successive elements.
- Strided memory access and hardware scatter/gather.
 - The program accesses elements of a vector located at fixed intervals.



Vector processors - Pros

- Fast.
- Easy to use.
- Vectorizing compilers are good at identifying code to exploit.
- Compilers also can provide information about code that cannot be vectorized.
 - Helps the programmer re-evaluate code.
- High memory bandwidth.
- Uses every item in a cache line.



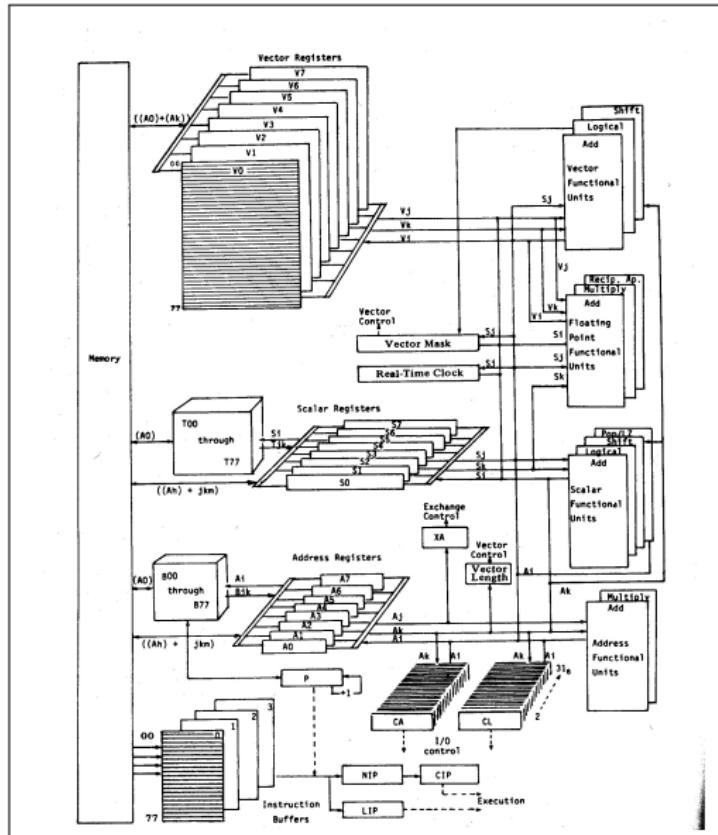
Vector processors - Cons

- They don't handle irregular data structures as well as other parallel architectures.
- A very finite limit to their ability to handle ever larger problems. (scalability)



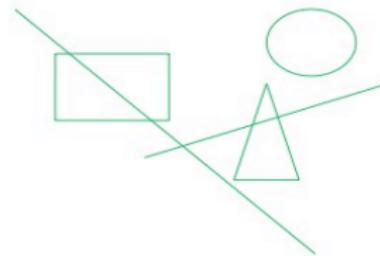
The Cray-1 Vector Computer:

- First vector machine (1975)
- \$8.86 million
- appx 140 MFlops, for weather calculation!!
- load a lot of data into memory, perform a lot of ops on that data
- Freon liquid cooling
- 12 functional units (address, scalar, vector, and floating point)



Graphics Processing Units (GPU)

- Real time graphics application programming interfaces or API's use points, lines, and triangles to internally represent the surface of an object.



GPUs

- A graphics processing pipeline converts the internal representation into an array of pixels that can be sent to a computer screen.
- Several stages of this pipeline (called shader functions) are programmable.
 - Typically just a few lines of C code.



GPUs

- Shader functions are also implicitly parallel, since they can be applied to multiple elements in the graphics stream.
- GPU's can often optimize performance by using SIMD parallelism.
- The current generation of GPU's use SIMD parallelism.
 - Although they are not pure SIMD systems.



NVIDIA GPU GF100 High-Level Block Diagram (2010)

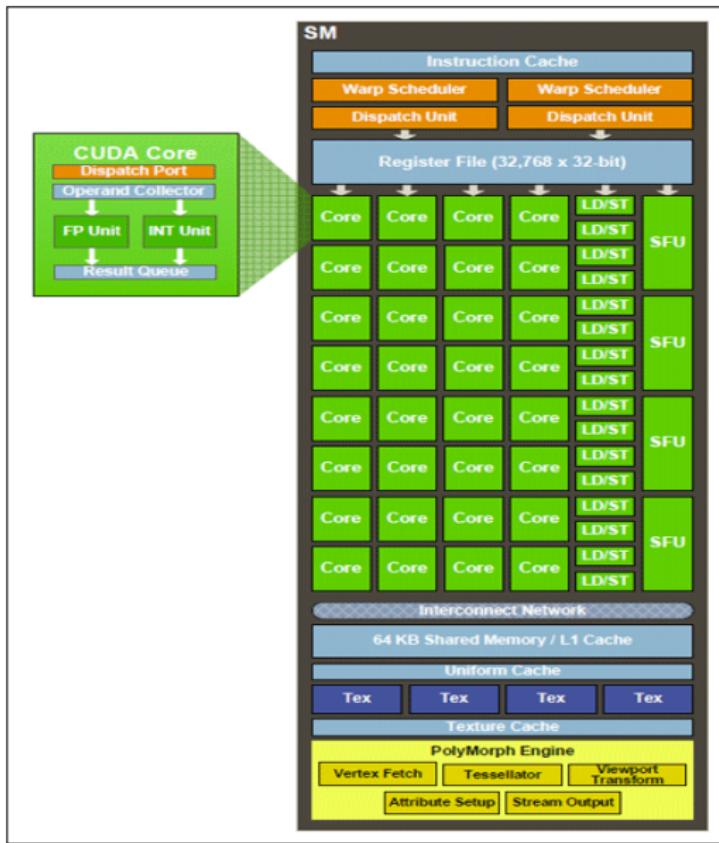
- CPU is called the host and the cores in the GPU are called devices
- 4 "GPC" clusters
- Many SM (stream multiprocessors) each with SPs
- 512 CUDA stream processors (SPs) or cores
- SIMT (single instr. multiple thread)



Source: <http://hothardware.com/Articles/NVIDIA-GF100-Architecture-and-Feature-Preview>

NVIDIA GPU

- each SM core in each GPC is comprised of 32 CUDA cores
- 48/16KB of shared memory (3 x that of GT200),
- 16/48KB of L1 (there is no L1 cache on GT200),

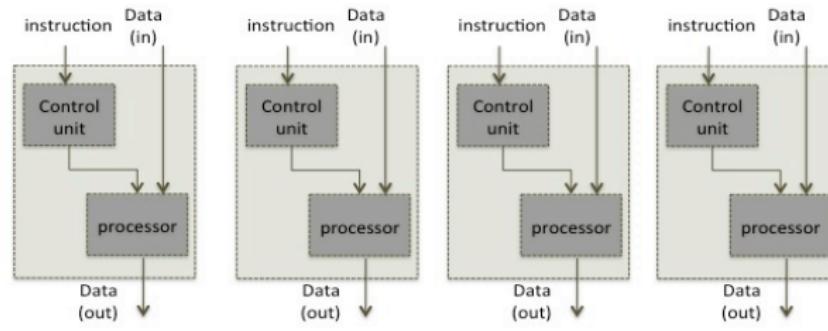


MIMD

- Supports multiple simultaneous instruction streams operating on multiple data streams.
- Typically consist of a collection of fully independent processing units or cores, each of which has its own control unit and its own ALU.



Single Instruction Multiple Data



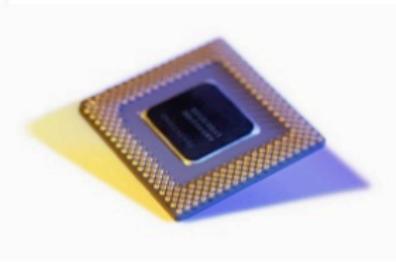
Shared Memory System (1)

- A collection of autonomous processors is connected to a memory system via an interconnection network.
- Each processor can access each memory location.
- The processors usually communicate implicitly by accessing shared data structures.



Shared Memory System (2)

- Most widely available shared memory systems use one or more multicore processors.
 - (multiple CPU's or cores on a single chip)



Shared Memory System

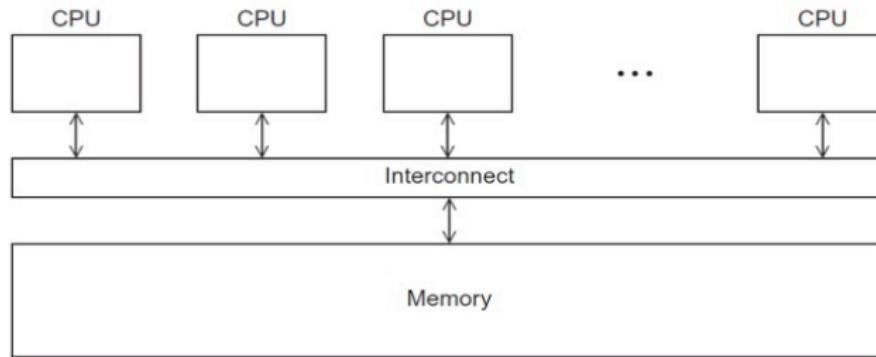
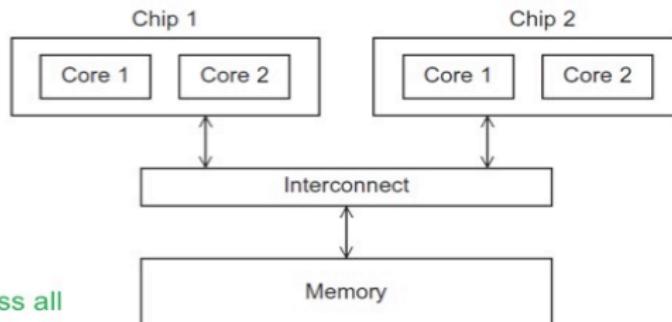


Figure 2.3

UMA multicore system

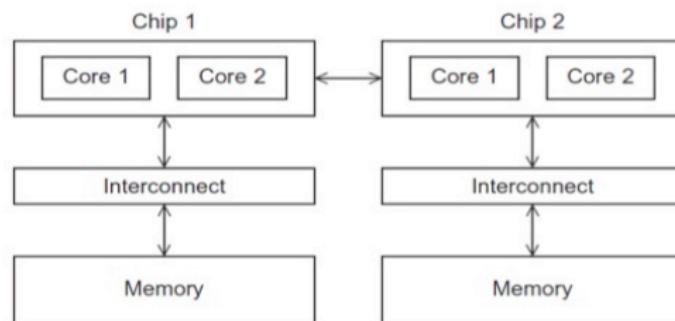


Time to access all
the memory locations
will be the same for
all the cores.

Figure 2.5



NUMA multicore system



A memory location a core is directly connected to can be accessed faster than a memory location that must be accessed through another chip.

Figure 2.6



Distributed Memory System

- Clusters (most popular)
 - A collection of commodity systems.
 - Connected by a commodity interconnection network.
- Nodes of a cluster are individual computations units joined by a communication network.

a.k.a. hybrid systems



Distributed Memory System

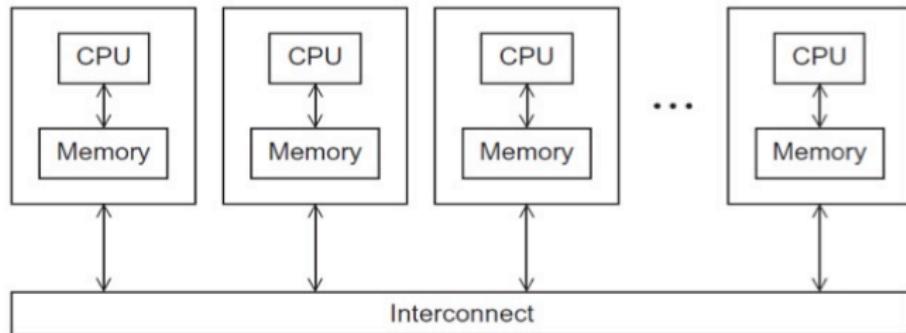
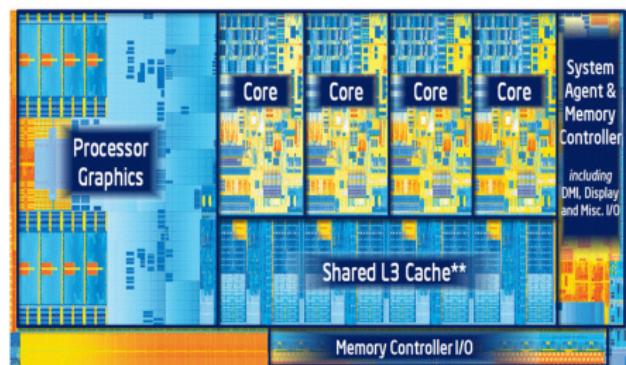


Figure 2.4

Mac Book Pro - Intel Core

- Intel Core i7, Z77 chipset
- 4 cores, 8 hyperthreads
- 32 + 32 KB L1 cache for data and instructions (per core)
- 256 KByte L2 cache (per core)
- 8 MB L3 cache (split up between cores and GPU)



Interconnection networks

- Affects performance of both distributed and shared memory systems.
- Two categories:
 - Shared memory interconnects
 - Distributed memory interconnects



Shared memory interconnects

- Bus interconnect
 - A collection of parallel communication wires together with some hardware that controls access to the bus.
 - Communication wires are shared by the devices that are connected to it.
 - As the number of devices connected to the bus increases, contention for use of the bus increases, and performance decreases.



Shared memory interconnects

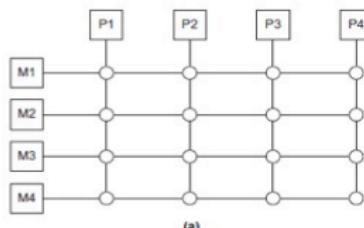
- Switched interconnect
 - Uses switches to control the routing of data among the connected devices.
- Crossbar –
 - Allows simultaneous communication among different devices.
 - Faster than buses.
 - But the cost of the switches and links is relatively high.



Figure 2.7

(a)

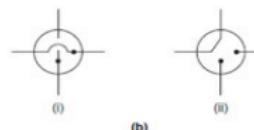
A crossbar switch connecting 4 processors (P_i) and 4 memory modules (M_j)



(a)

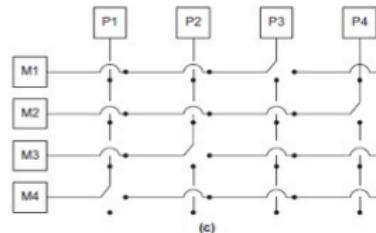
(b)

Configuration of internal switches in a crossbar



(b)

(c) Simultaneous memory accesses by the processors



(c)



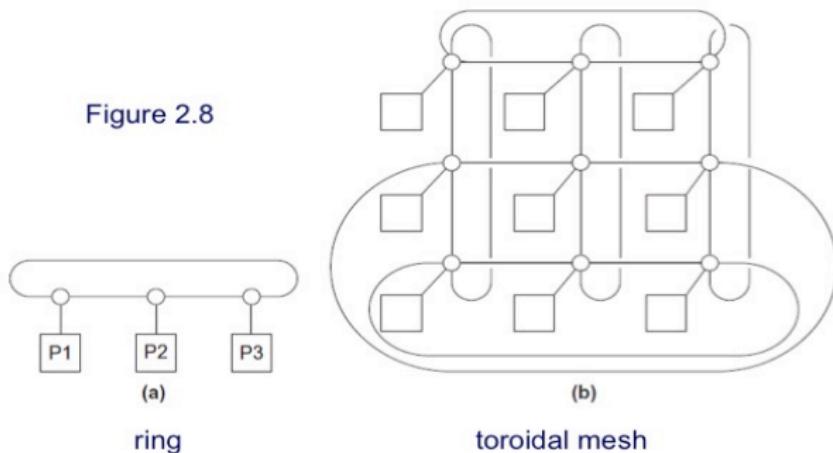
Distributed memory interconnects

- Two groups
 - Direct interconnect
 - Each switch is directly connected to a processor memory pair, and the switches are connected to each other.
 - Indirect interconnect
 - Switches may not be directly connected to a processor.



Direct interconnect

Figure 2.8



Bisection width

- A measure of “number of simultaneous communications” or “connectivity”.
- How many simultaneous communications can take place “across the divide” between the halves?



Two bisections of a ring

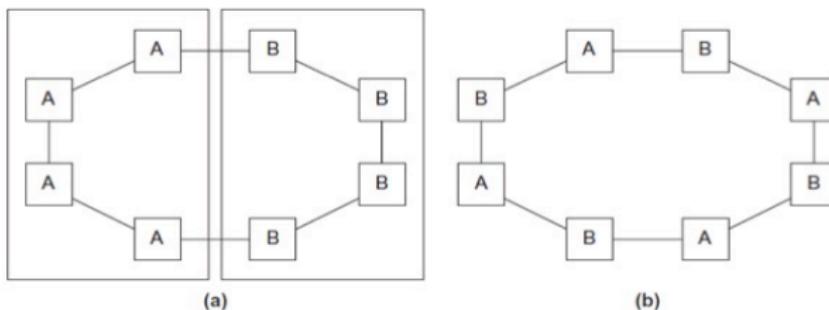


Figure 2.9

A bisection of a toroidal mesh

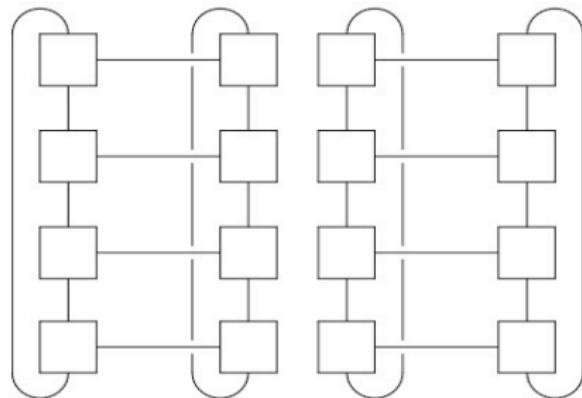


Figure 2.10

Definitions

- Bandwidth
 - The rate at which a link can transmit data.
 - Usually given in megabits or megabytes per second.

- Bisection bandwidth
 - A measure of network quality.
 - Instead of counting the number of links joining the halves, it sums the bandwidth of the links.



Fully connected network

- Each switch is directly connected to every other switch.

impractical

$$\text{bisection width} = p^2/4$$

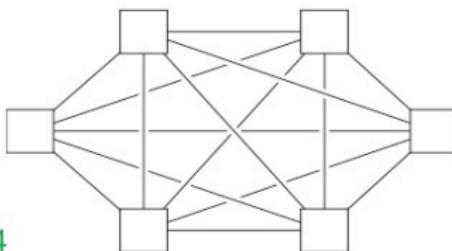


Figure 2.11

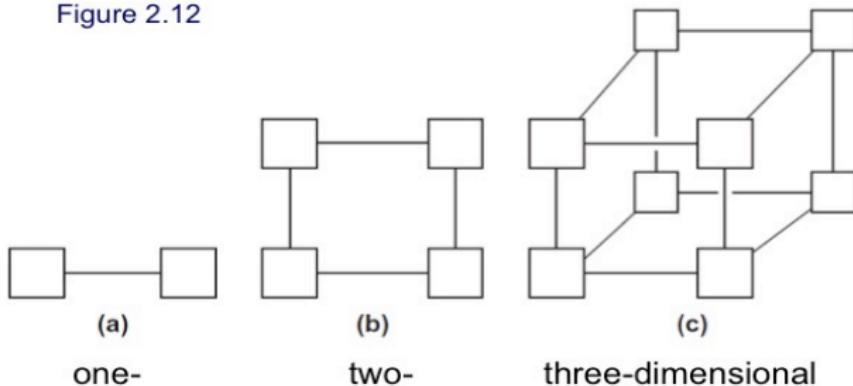
Hypercube

- Highly connected direct interconnect.
- Built inductively:
 - A one-dimensional hypercube is a fully-connected system with two processors.
 - A two-dimensional hypercube is built from two one-dimensional hypercubes by joining “corresponding” switches.
 - Similarly a three-dimensional hypercube is built from two two-dimensional hypercubes.



Hypercubes

Figure 2.12



Indirect interconnects

- Simple examples of indirect networks:
 - Crossbar
 - Omega network
- Often shown with unidirectional links and a collection of processors, each of which has an outgoing and an incoming link, and a switching network.



A generic indirect network

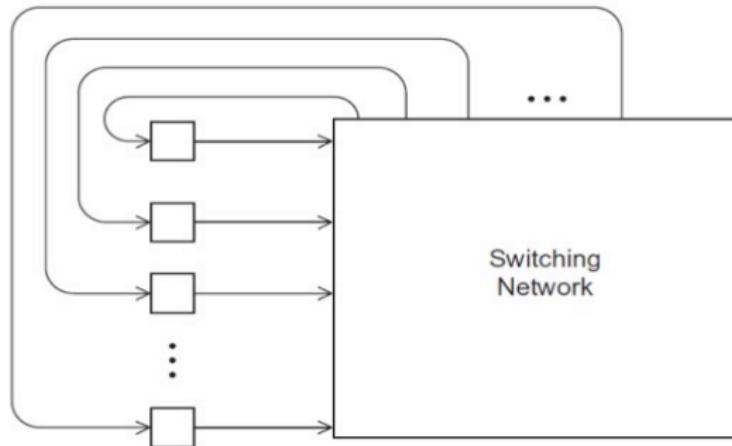


Figure 2.13

Crossbar interconnect for distributed memory

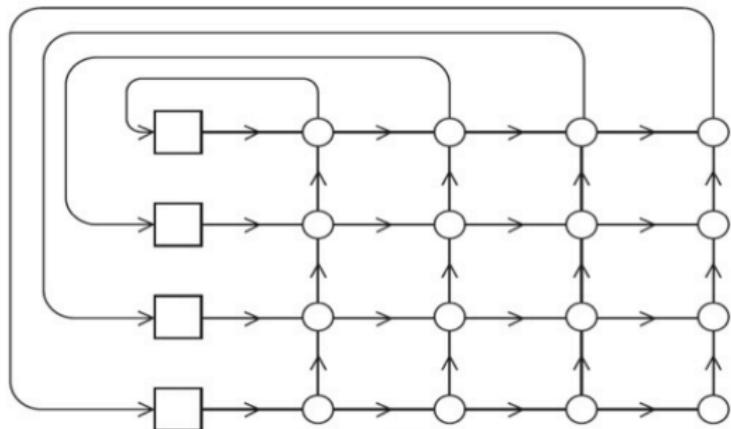


Figure 2.14

An omega network

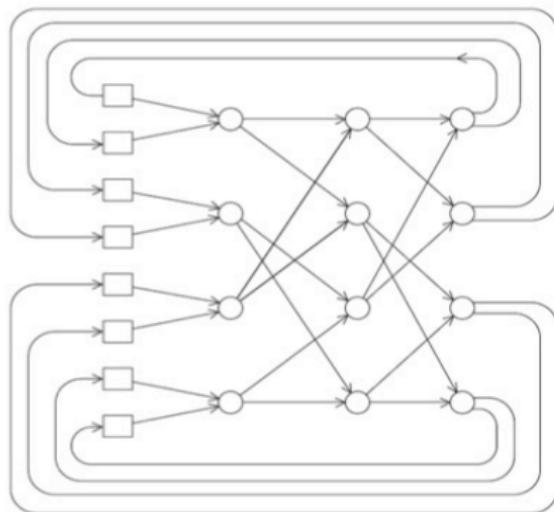


Figure 2.15

A switch in an omega network

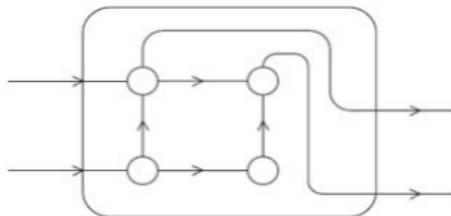


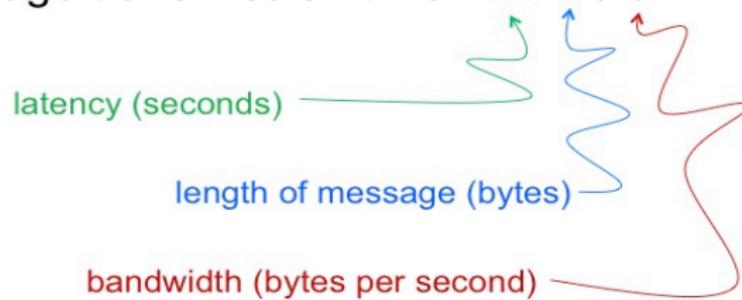
Figure 2.16

More definitions

- Any time data is transmitted, we're interested in how long it will take for the data to reach its destination.
- **Latency**
 - The time that elapses between the source's beginning to transmit the data and the destination's starting to receive the first byte.
- **Bandwidth**
 - The rate at which the destination receives data after it has started to receive the first byte.

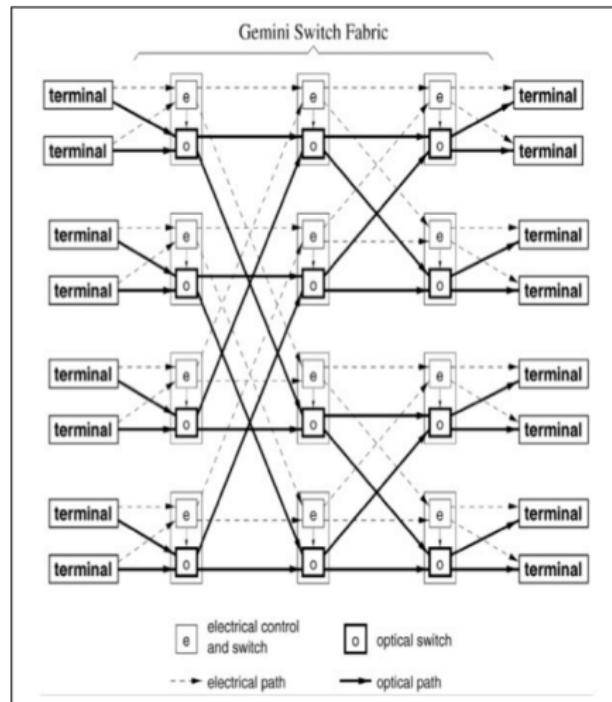


Message transmission time = $l + n / b$



ORNL Titan Supercomputer - Jaguar upgrade

- 38,400-processors, 307,200 CPU cores
- 20-petaflop [AMD Opteron 6200]
- Cray Gemini Interconnect
 - processor-to-processor
 - Optical network of 2x2 switches
 - Banyan: $O(N \log N)$



Source: <http://www.extremetech.com/extreme/99413-titan-supercomputer-38400-processor-20-petaflop-successor-to-jaguar>

Cache coherence

- Programmers have no control over caches and when they get updated.

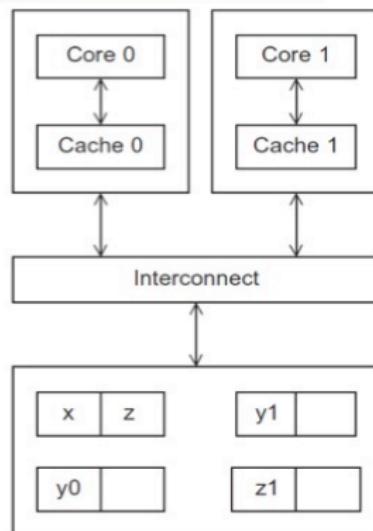


Figure 2.17

A shared memory system with two cores and two caches



Cache coherence

y0 privately owned by Core 0

y1 and z1 privately owned by Core 1

x = 2; /* shared variable */

Time	Core 0	Core 1
0	y0 = x;	y1 = 3*x;
1	x = 7;	Statement(s) not involving x
2	Statement(s) not involving x	z1 = 4*x;

y0 eventually ends up = 2

y1 eventually ends up = 6

z1 = ???



Snooping Cache Coherence

- The cores share a bus .
- Any signal transmitted on the bus can be “seen” by all cores connected to the bus.
- When core 0 updates the copy of x stored in its cache it also broadcasts this information across the bus.
- If core 1 is “snooping” the bus, it will see that x has been updated and it can mark its copy of x as invalid.



Directory Based Cache Coherence

- Uses a data structure called a directory that stores the status of each cache line.
- When a variable is updated, the directory is consulted, and the cache controllers of the cores that have that variable's cache line in their caches are invalidated.

