**Basic Computational Biology** 

### High Performance Computing Technology(1)

### Introduction to parallel computing and systems

M. Sato

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## What is HPC ?

- Today's science (domain science) is driven by three elements
  - Experiment
  - Theory
  - Computation (Simulation)
- In many of these problems, computation performance and capacity are required to be larger and larger
  - Floating point operation speed
  - Memory capacity (amount)
  - Memory bandwidth (memory speed)
  - Network bandwidth (network speed)
  - Disk (2nd storage) capacity
- "High Performance" does not mean only the speed but also capacity and bandwidth

### **Computational science**

- Large-scale simulations using supercomputers
- Critical and cutting-edge methodology in all of science and engineering disciplines
- The third "pillar" in modern science and technology



### What computational science can do ...

- To explore complex phenomenon which cannot be solved by "paper and pencil"
  - Particle physics to explore origin of materials
  - Phenomenon caused by aggregation of DNAs and protein
- To explore phenomenon which cannot be solve by experiment
  - Origin of universe
  - Global Warming of the earth
- To analyze a large set of data "big data"
  - Genome informatics
- To reduce the cost by replacing expensive experiments
  - car crash Simulation
  - CFD to design air craft

First principal method: computer simulation based on only computation without "experimental parameters. But it may require a huge computations



DNA



## "First principal computation"...

Schrödinger equation

$$\mathbf{H}\psi = E\psi \quad i\hbar\frac{\partial\psi}{\partial t} = \left[-\frac{\hbar^2}{2m}\frac{d^2}{dx^2} + V(x)\right]\psi$$

 "first principle calculation(computation)" in computational material



### How to make Computer fast ...

- Metric of speed of computation: arithmetic operations (floating point) per second
- MFLOPS: <u>Millions of FL</u>oating Point <u>OP</u>eration<u>S</u>.
- GFLOPS: 10<sup>9</sup> ops, TFLOPS: 10<sup>12</sup> ops, PFLOPS: 10<sup>15</sup> pos, Exa
- ① By making electric circuit work fast
   Increasing clock speed (Frequency of processors used in PC: 2~3GHz)
  - Using fast transistor



## History of hardware of supercomputers

- 1983: 1 GFLOPS, 1996:1 TFLOPS...
- Before 1990's, the main stream of "supercomputer" was vector supercomputer
- Rapid progress of microprocessor (all components in a chip) used for PC --- "killer micro"
  - Moore's Low: integration (density) of transistor increase double per 1.5 year
  - 4004(first microproessor、1971、750KHz) 8008(1972、500KHz、 Intel) 8080(1974、2MHz、Intel)
  - Pentium 4 (2000、~3.2GHz)

### Clock speed increased from 1MHz to 1GHz in the last three decades<sub>8</sub>

### To make computer fast ...

- By good mechanisms (architecture) in computer
  - mechanism to execute many instruction at a time (in one clock ...)
  - Vector supercomputer: a computer with computing unit to execute vector computation frequently used in scientific computation

(1980's)



To make computer fast ...

- ③ by using many computer at a time
  - Parallel computers, parallel processing …
  - This is a main stream in supercomputer !
  - You can find 2 or 3 processors in a PC or "smart phone"!

### Moore's Law re-interpreted

- Progress of clock speed stops after 2000's
- Still increasing the number of transistors
- Multicore
  - Core (computer) in onechip
  - double in the number of cores every 18 months



### TOP 500 List: How to measure (rank) performance of supercomputers http://www.top500.org/

- Ranked by the performance of benchmark program "LINPACK"
  - LINPACK solves a huge size of linear equations
  - the size is more than 10 millions
- Different from the performance of "real" applications
  - It does not necessarily reflect the performance of "real" applications
- The power consumption is indicated since 2008
  - The power saving is import now !

### TOP500:全世界のスパコンランキング500位



Very simple example of parallel computing for high performance



### Shared memory multi-processor system



- Multiple CPUs share main memory
- Threads executed in each core(CPU) communicate with each other by accessing shared data in main memory.
- Enterprise Server
   SMP Multi-core processors

### Distributed memory multi-processor



- System with several computer of CPU and memory, connected by network.
- Thread executed in each computer communicate with each other by exchanging data (message) via network.夕
- PC ClusterAMP Multi-core processor

## 京コンピュータ "The K computer"



### Facts of the K computer

- The number of racks (boxes) 864
- the number of chips 82,944
- The number of cores (computers) 663,552
- Linpack perf
   10.51PF
   (Power 12.66MW)
   2011/11月



## Amdahl's low

Question: How much do parallel computers became fast by increasing the number of processors???

> ジーン・アムダール(Gene Amdahl、 1922年11月16日 - )は、アメリカ人の コンピュータアーキテクトで、企業家あ る。彼の業績はIBMおよび彼の創設し た会社(特にアムダール社)における、 メインフレームの設計である。並列コン ピューティングの基本的な理論としてア ムダールの法則がよく知られている。 (wikipediaより)



### Speedup by parallel computing: "Amdahl's low"

### Amdahl's low

- Suppose execution time of sequential part  $T_1$ , ratio of sequential part a, execution time by parallel computing using p processors  $T_p$ is (no more than)  $T_p = a * T_1 + (1-a) * T_1/p$
- Since some part must be executed sequentially, speedup is limited by the sequential part.



### Breaking "Amdahl's low"

"Gustafson's low": what about performance of real apps?

- The fraction of parallel part often depends on the size of problem
- For example, n-times larger problem to be solve by n-times larger parallel computers.
- Weak scaling Scaling with constant size per processor ← in the case of large scale scientific applications
- Strong scaling Scaling with constant size problem  $\leftarrow$  We need fast one-processor.



How different between the K computer and your PC?

- The processors (computer) used are almost the same!
  - Even slow clock for the K computer, but some enhancement in computing unit.
- The K computer consists of many "processors"
  - 80,000 chip、0.64 M cores
  - Fast network between processors is required!
- The programmer is forced to make parallel program to make use of many processors
  - The program running on the PC (sequential program) does not run fast !

## Parallel computing

- For efficient parallel processing, certain "granularity" of parallel processing unit and enough degree of parallelisms are necessary
- Ordinary (non-scientific) applications are not sufficient to satisfy these conditions naturally
  - ex. "Word" or "Excel" applications do not have parallelism nor large amount of computation in a second
- Various scientific computations satisfy these conditions, and there are much requirement of solving these problems (especially for high-end domain science)
- Large scale parallel processing is naturally getting along with HPC
- So many numerical algorithms have been developed for scientific computation which is enable on parallel systems
- In many cases, matrix computation is essential, but direct solution is more effective in some cases

# Why parallelization needs? 4 times speedup by using 4 cores!



## Parallel Processing and Distributed Processing

- parallel processing is defined as a technology to process/compute faster by using many processors simultaneously
  - HPC(High Performance Computing)
    - scientific simulation "supercomputing"
  - HTC (High Throughput Computing)
    - processing a huge amount of data "big data"
- Distributed processing is referred as a technology to process/compute by using many processors, but it federate several functions executed in different computers to provide high-level services.
  - Distributed objects ...
  - RMI , J2EE, Jini...

### Some terminologies

- Node A standalone "computer in a box". Usually comprised of multiple CPUs/processors/cores. Nodes are networked together to comprise a parallel system.
- Task A logically discrete section of computational work. A parallel program consists of multiple tasks running on multiple processors.
- Communications Parallel tasks typically need to exchange data. There are several ways this can be accomplished, such as through a shared memory bus or over a network.
- Synchronization The coordination of parallel tasks in real time, very often associated with communications. Often implemented by establishing a synchronization point with an applications where a task may not proceed further until another task(s) reaches the same or logically equivalent point.

## Some terminologies

- Granularity in parallel computing, granularity is a qualitative measure of the ratio of computation to communication.
  - Coarse : relatively large amount of computational work are done between communication events
  - Fine: relatively samll amount of computational work are done between communication events
- Parallel overhead The amount of time required to coordinate parallel tasks, as opposed to doing useful work.
   Parallel overhead can include factors such as:
  - Task start-up time
  - Synchronization
  - Data communications
  - Software overhead imposed by parallel compiler, libs, tools,
     ...
  - Task terminations



### Overhead of parallel execution



### Some terminologies

- Scalability Refers to a parallel system's (hardware and/or software) ability to demonstrate a proportionate increase in parallel speedup with the addition of more processors. Factors that contribute to scalability include:
  - Hardware particularly memory-cpu bandwidth and network communications
  - Application algorithm
  - Parallel overhead related
  - Characteristics of your coding and apps.

# Metric of Performance of Parallel Systems

#### Speed up

- T : execution time by 1 processor
- *T(p)* : execution time by *p* processors
- s(p)=T/T(p)
   s(p)を: speedup by processor p. if s(p) is more than 1, the speed of computation increases
- Ideally it should be s(p)=p (p台のプロセッサを投入した結果、p倍の速度が得られた)



# Metric of Performance of Parallel Systems

- Efficiency
  - Speedup is not useful since s(p) depends on p
  - Suppose [s(p)=p is ideal], this metric is defined as how much this ideal is archived.
  - e(p)=s(p)/p
     e(p) does not depend on p. It is good if it is close to 1



### Data Parallel Model

- The data parallel models demonstrates the followings:
  - Most of the parallel work focuses on performing operations on a data sets. The data set is typically organized into common structure, such as an array or cube.
  - A set of task work collectively

on the same data structure, however, each task works on different partition of the same data structure.

 Tasks perform the same operation on their partition of work



# Example of data parallel model

- domain decomposition
  - Divide the space of simulation into uniform grids
  - Perform the same computation on each gird, sometimes with interaction of neighbor



grid for computational unit

example:

```
for(t=0; t < T; t++){
    for(i=0; i < N; i++)
        a[i] = b[i-1] + 2*b[i] + b[i+1];
    for(i=0; i < N; i++)
        b[i] = a[i];
}
</pre>
```



simulation space

## Simple Heat Equation

- Most problems in parallel computing require communication among the tasks. A number of common problem require communications "neibhbor" task. (stencil computations)
- A finite difference scheme is employed to solve the heat equations numerically on a square regions.
- For the fully explicit problem, a time stepping algorithm is used. The element of a 2-dimensional array represent the temperature at the point on the square.

/+1

x

U x+1,y

$$U_{x,y} = U_{x,y}$$
  
+  $C_x * (U_{x+1,y} + U_{x-1,y} - 2 * U_{xy})$   
+  $C_y * (U_{x,y+1} + U_{x,y-1} - 2 * U_{x,y})$   
U x.1.y U x  
U x.1.y U x



## Simple Heat Equation

- The entire array is partitioned and distributed as subarray to all task. Each task owns a portion of the total array.
  - send slave read of u1 to neighbor processor
  - receive u1
  - compute u2 at each processor
  - update u1 with u2
  - repeat the above computation until the condition is satisfied.

```
do iy = 2, ny-1
do ix = 2, nx-1
u2(ix,iy) = u1(ix,iy)+
    cx*(u1(ix+1),y)+u1(ix+1,iy)-2*u1(ix,iy))+
    cy*(u1(ix,iy+1)+u1(ix,iy-1)-2*(ix,iy))
end do
end do
end do
```



# Pipeline

- Breaking a task into steps performed by different processors unit, with inputs streams through, much like assemble lines
- Example: signal processing



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# master/worker parallel processing

- one master processor and several worker processors
- A pool of work in master processor.
- master pick up one work to send the work to a worker.
- When worker finish the given work, then it return the result and receive next work

```
master:: worker:: 
// give a job to each worker while(1){
    // receive a worker's result
    // give the next job to that worker
} 
worker:: 
while(1){
    // receive a job from master
    // process the job
    // send the result to master
    }
```

# master/worker parallel processing

 It is effective parallel processing when each work have different load --> load balancing



# Load Balancing

- Load Balancing refers to the practice of distributing work among tasks so that all tasks kept busy all of the time. It can be considered a minimization of task idle time.
- Load balancing is important to parallel programs for performance. For example, if all tasks are subject to a barrier sync point, the slowest task will determine the overall performance.
- How to achieve load balance:
  - Equally partition the work each tasks receive.
  - Use dynamic work assignment
    - Master-Worker



### Example: Molecular Dynamics with cut-off radius

- MD (Molecular Dynamics)
  - Compute interaction between P particles in ndimensional space.
  - Interaction may be force between particles
  - If the force is effective only within near fields, cut-off distance can be assumed.
- To save the computations, only computation within the cut-off radius should be done, not all-to-all interaction,
  - Space is divided into cell by "domain decomposition" assigned to each node, and each node computes particles within the assigned cell
     ⇒ If the size of cell is larger than the cut-off radius, nodes may communicate only with nodes of neighbor cells (cell mapping method)



## Example: Molecular Dynamics with cut-off

- Particles moves by the force of interaction from other particles, as steps go. As a result, it may happens that many particles moves into a certain cell. (condense)
- In the case that cell is assigned to nodes in oneto-one manner, load imbalance may occur.
- In order to keep load balance, the number of particles computed by nodes should be balanced rather than the number of cells.
- Methods:
  - Method 1) Periodically, the number of cells are reassigned (adjusted) according to the density of particles in the cell (the number of particle/cell)
  - Method 2) If the number of cells is far more than the number of nodes, use cyclic mapping rather than block mapping.
  - Method 3) Use particle mapping, not cell mapping.



## Example: Molecular Dynamics with cut-off

Method 1)

To re-assign cells to nodes, a large amount of data should be exchanged (needs much comm) Since assignment will be irregular, the communication pattern is not neighbor communication.

Method 2)

Cyclic mapping is a simple way to take a good load balance. But, the communication pattern is not neighbor communication.

Method 3)

To keep track which nodes each particle is assigned to, the table to manage the index table of the assignment between particles and nodes, resulting a complicated and expensive computation and communications.

 $\Rightarrow$  No best solution for all cases.

- Depends on the characteristics of phenomenon to be solved (how particle behaves, or what potential force.).
- It may important to keep load balancing in the case of heavily load imbalance.