## Parallel Processing & Distributed Systems

## Lectured by: Phạm Trần Vũ Prepared by: Thoại Nam



- □ Two lectures per week (90 minutes each)
  - Tuesday: 10:00 11:35
  - Thursday: 8:15 9:50
- □ References
  - Scalable Parallel Computing: Technology, Architecture, Programming, Kai Hwang & Zhiwei Xu, McGRAW-HILL, 1997.(\*)
  - Parallel Computing theory and practice, Michael J. Quinn, McGRAW-HILL, 1994.(\*)
  - Parallel Programming: Techniques and Applications Using Networked Workstations and Parallel Computers, Barry Wilkinson and MiChael Allen, Second Edition, Prentice Hall, 2005.
  - Distributed Systems: Concepts and Design, George Coulouris, Jean Dillimore, Tim Kindberg, Addison-Wesley, 2005.(\*)
  - Distributed Algorithms, Nancy Lynch, Morgan Kaufmann, 1997.
  - *Distributed Operating Systems*, Andrew S. Tanenbaum, Prentice Hall, 1990.
  - *MPI*: http://www.mpi-forum.org/docs/docs.html
  - *PVM*: http://www.csm.ornl.gov/pvm/pvm\_home.html
  - The GRID2: Blueprint for a New Computing Infrastructure, Ian Foster and Carl Kesselman, Morgan Kaufmann 2004.
  - Grid Computing: Making the Global Infrastructure a Readlity, Fran Berman, Geoffgrey Fox and Tony Hey.

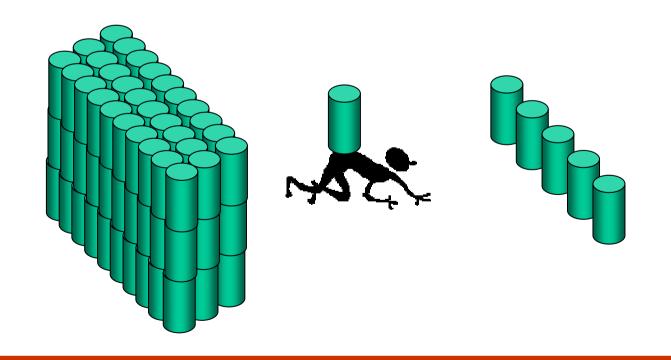


## Introduction

- What is parallel processing?
- Why do we use parallel processing?
- Applications
- Parallelism



- □ 1 CPU
- □ Simple
- □ Big problems???









A grand challenge problem is one that cannot be solved in a reasonable amount of time with today's computers

□ Ex:

- Modeling large DNA structures
- Global weather forecasting
- Modeling motion of astronomical bodies



□ Power processor

- 50 Hz -> 100 Hz -> 1 GHz -> 4 Ghz -> ... -> Upper bound?

Smart worker

- Better algorithms

□ Parallel processing



- $\Box$  The N<sup>2</sup> algorithm:
  - N bodies
  - N-1 forces to calculate for each bodies
  - N<sup>2</sup> calculations in total
  - After the new positions of the bodies are determined, the calculations must be repeated
- □ A galaxy:
  - 10<sup>7</sup> stars and so 10<sup>14</sup> calculations have to be repeated
  - Each calculation could be done in  $1\mu s$  ( $10^{-6}s$ )
  - It would take **10 years** for one iteration
  - But it only takes 1 day for one iteration with 3650 processors



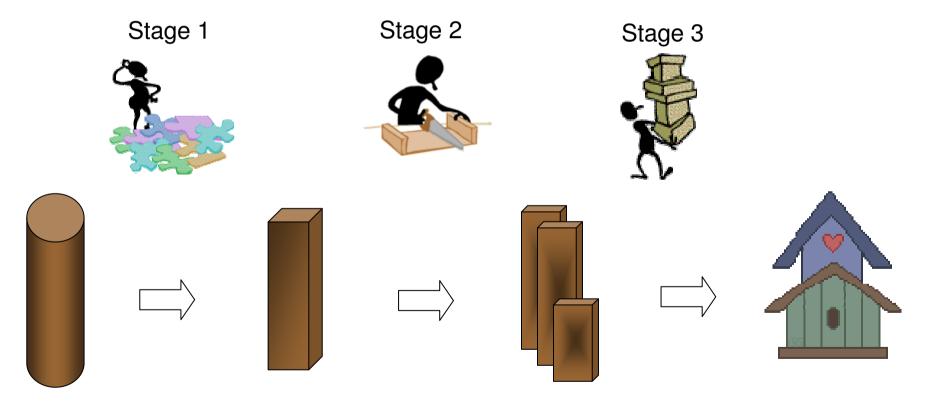
- Parallel processing
- Parallel computer
  - Multi-processor computer capable of parallel processing
- □ Throughput:
  - The throughput of a device is the number of results it produces per unit time.
- □ Speedup
  - S = Time(the most efficient sequential algorithm) / Time(parallel)

algorithm)

- □ Parallelism:
  - Pipeline
  - Data parallelism
  - Control parallelism

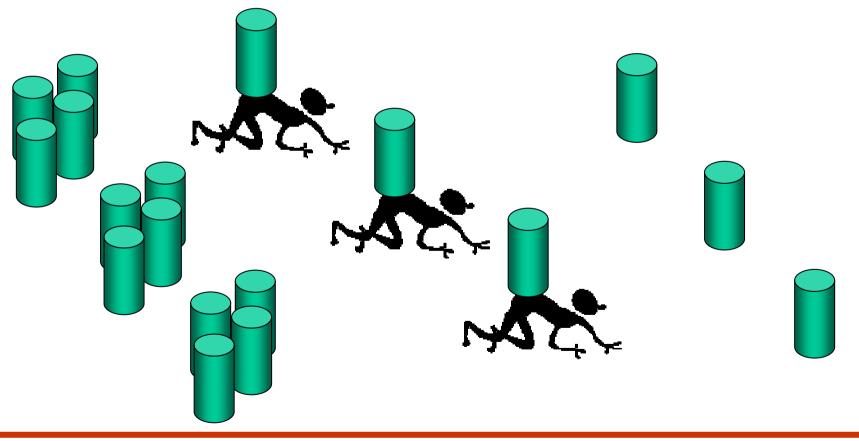


- □ A number of steps called **segments** or **stages**
- □ The output of one segment is the input of other segment

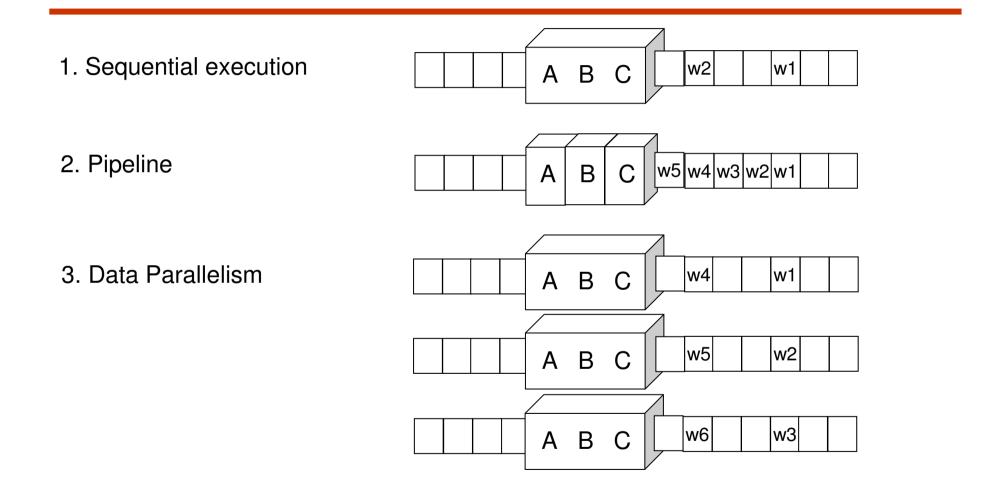




Applying the same operation simultaneously to elements of a data set









- □ Pipeline is a special case of control parallelism
- □ T(s): Sequential execution time

T(p): Pipeline execution time (with 3 stages)

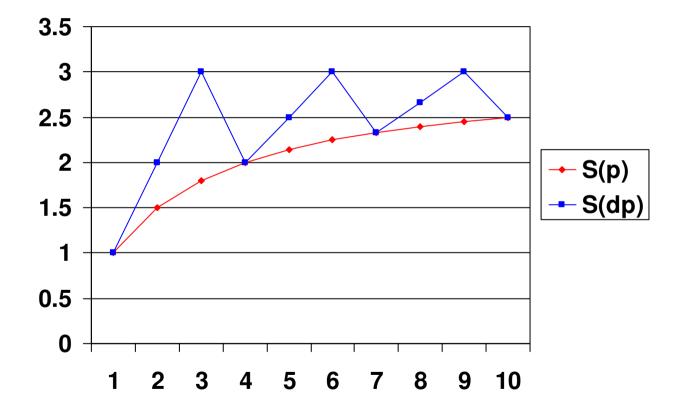
T(dp): Data-parallelism execution time (with 3 processors)

S(p): Speedup of pipeline

S(dp): Speedup of data parallelism

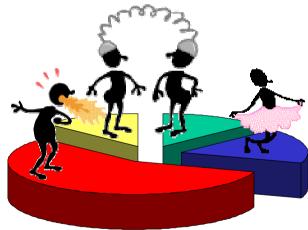
widget	1	2	3	4	5	6	7	8	9	10
T(s)	3	6	9	12	15	18	21	24	27	30
T(p)	3	4	5	6	7	8	9	10	11	12
T(dp)	3	3	3	6	6	6	9	9	9	12
S(p)	1	1+1/2	1+4/5	2	2+1/7	2+1/4	2+1/3	2+2/5	2+5/11	2+1/2
S(dp)	1	2	3	2	2+1/2	3	2+1/3	2+2/3	3	2+1/2







Applying different
operations to different
data elements
simultaneously





- An algorithm is scalable if the level of parallelism increases at least linearly with the problem size.
- An architecture is scalable if it continues to yield the same performance per processor, albeit used in large problem size, as the number of processors increases.
- Data-parallelism algorithms are more scalable than controlparallelism algorithms