Parallel Processing & Distributed Systems

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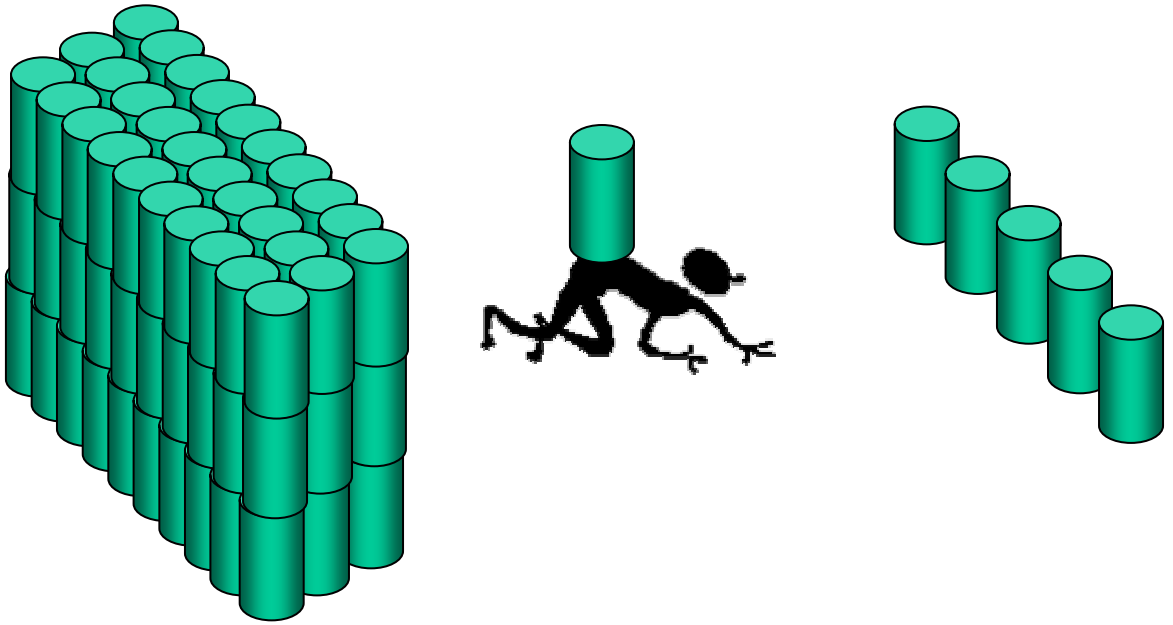
Chapter 1: Introduction

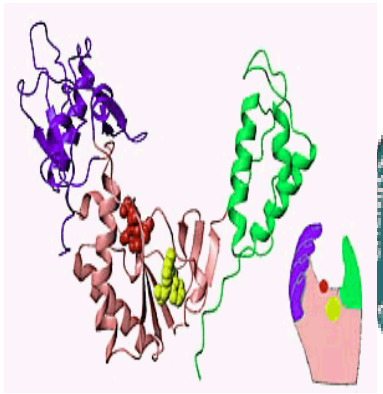
- Introduction
  - What is parallel processing?
  - Why do we use parallel processing?
- Applications
- Parallelism

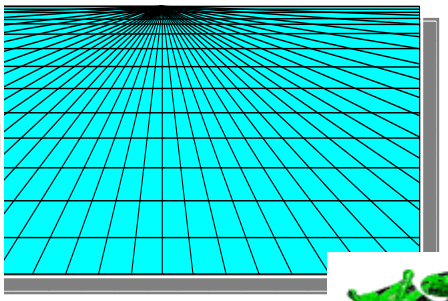
Sequential Processing

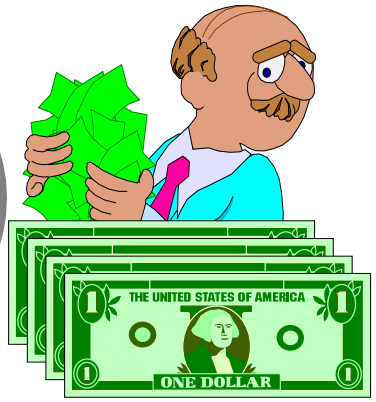
- 1 CPU
- Simple
- Big problems???

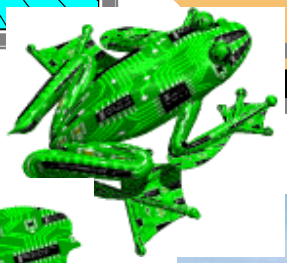


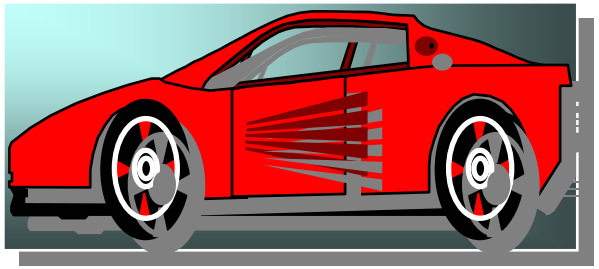
Application Demands

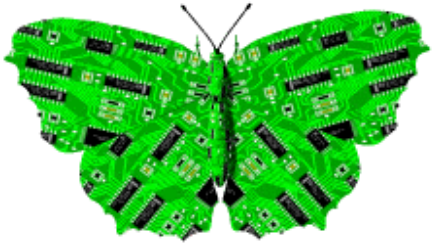




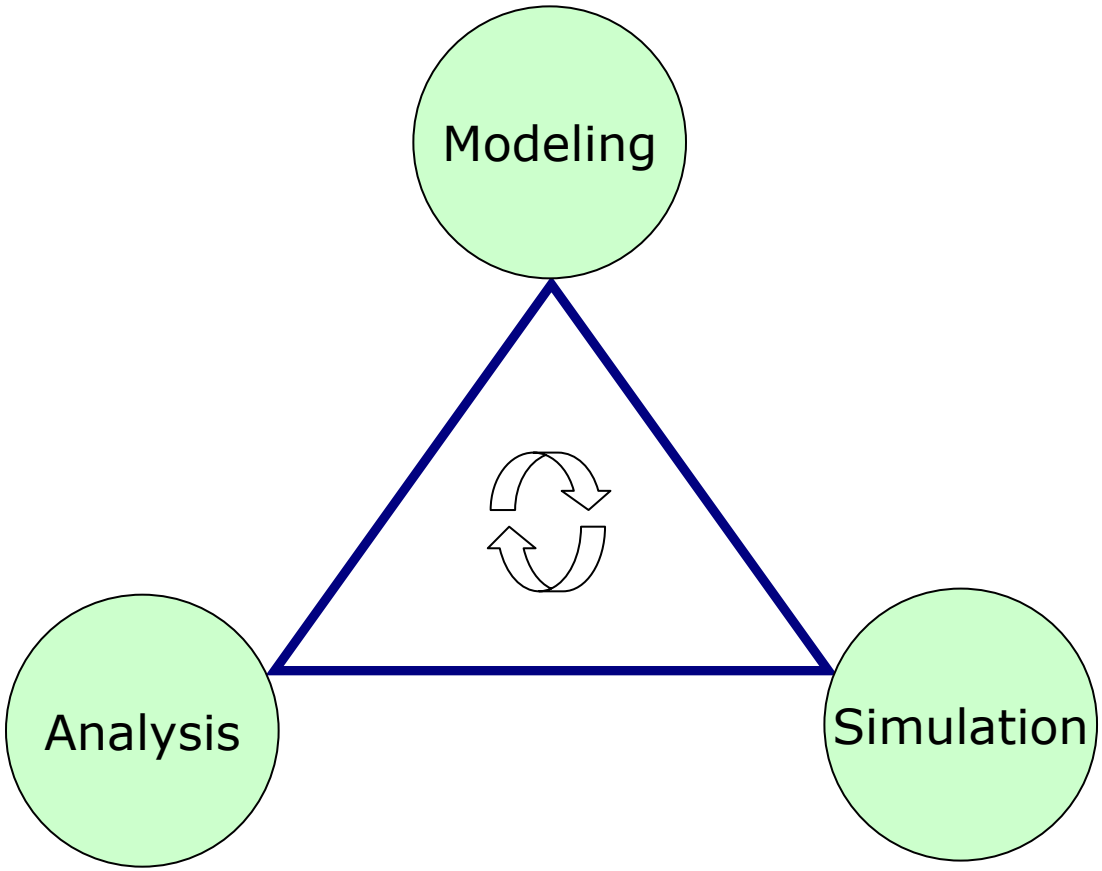








New Approach



Grand Challenge 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

Solutions

- Power processor
  - 50 Hz -> 100 Hz -> 1 GHz -> 4 Ghz -> ... -> Upper bound?
- Smart worker
  - Better algorithms
- Parallel processing

N-body

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

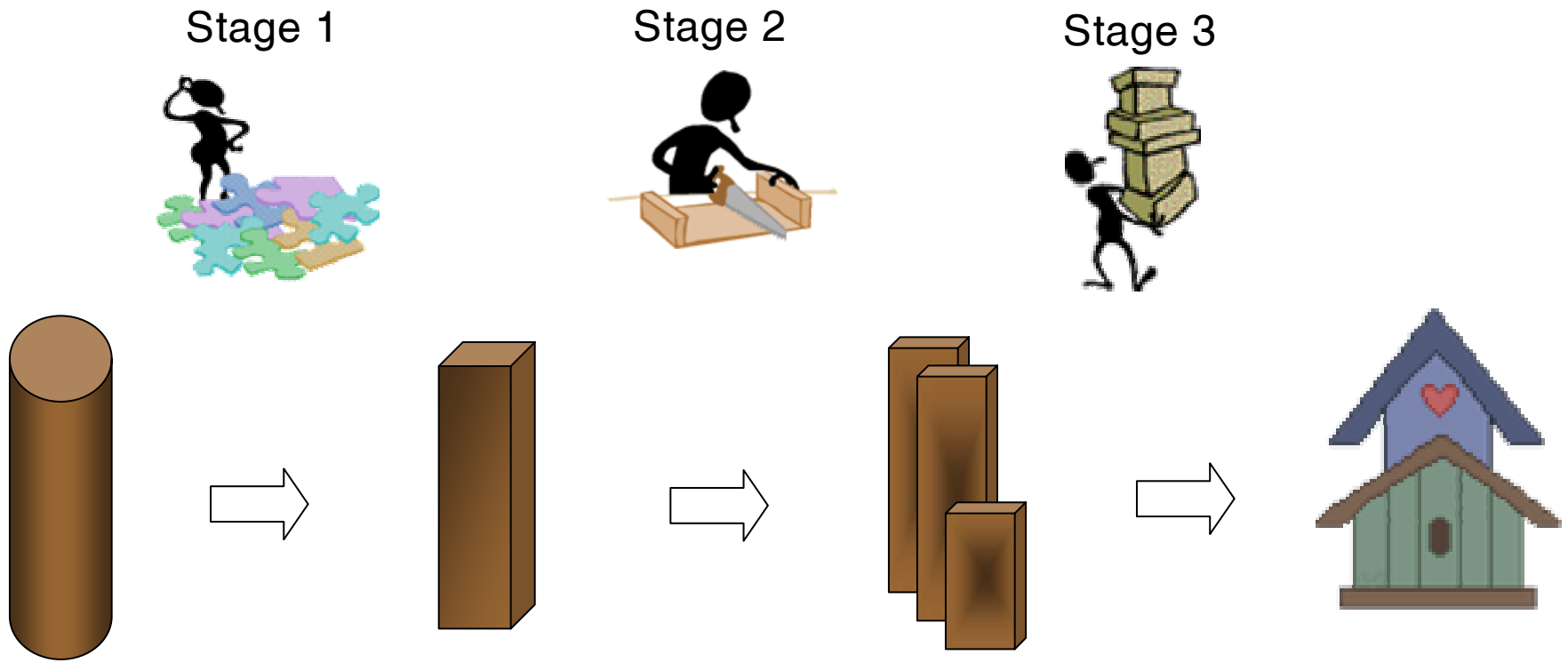
Parallel Processing Terminology

- 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

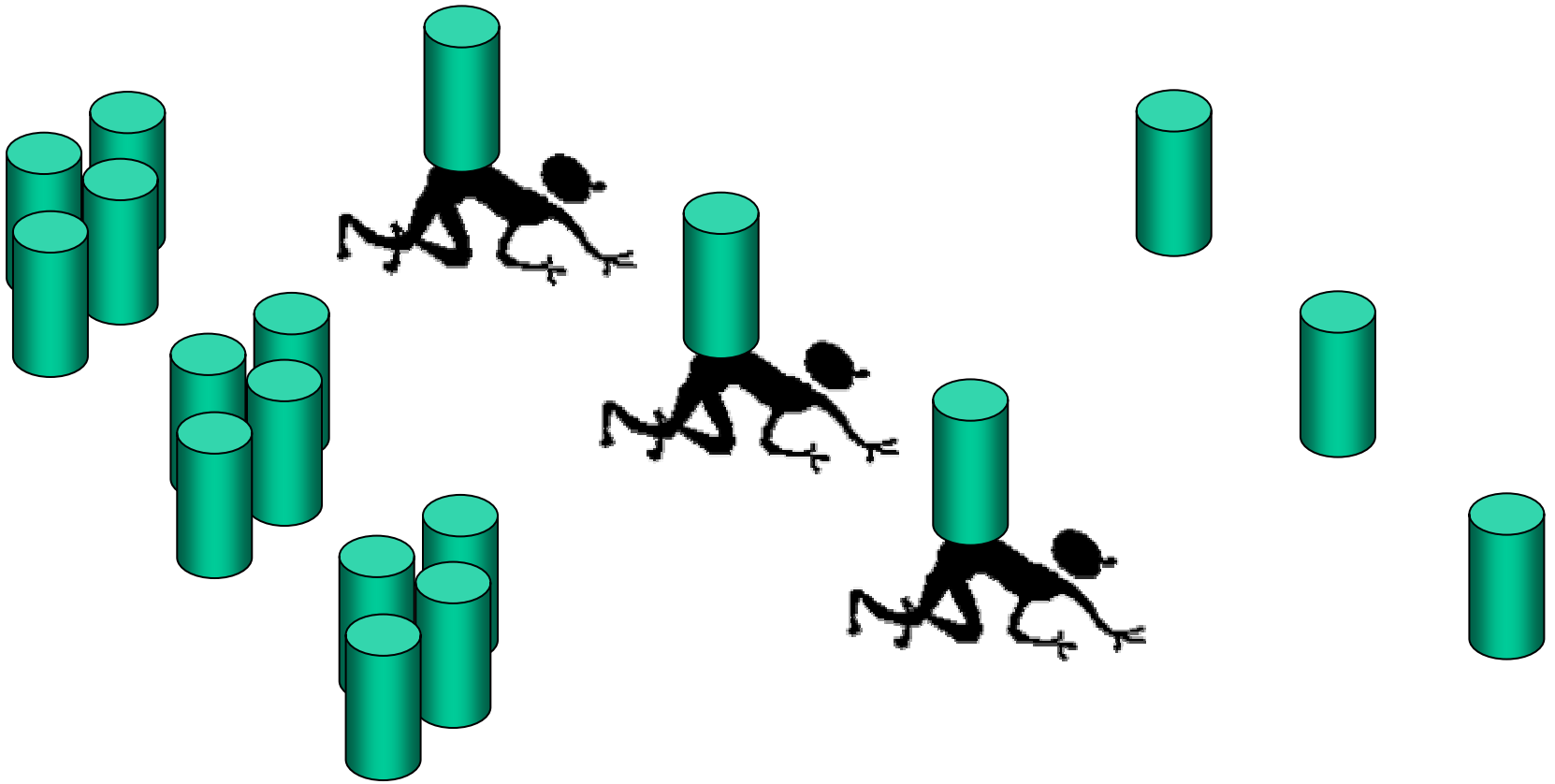
Pipeline

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



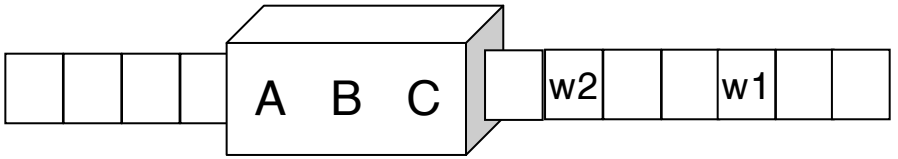
Data Parallelism

- Applying the same operation simultaneously to elements of a data set

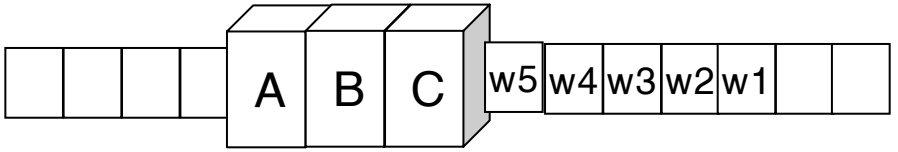


Pipeline & Data Parallelism

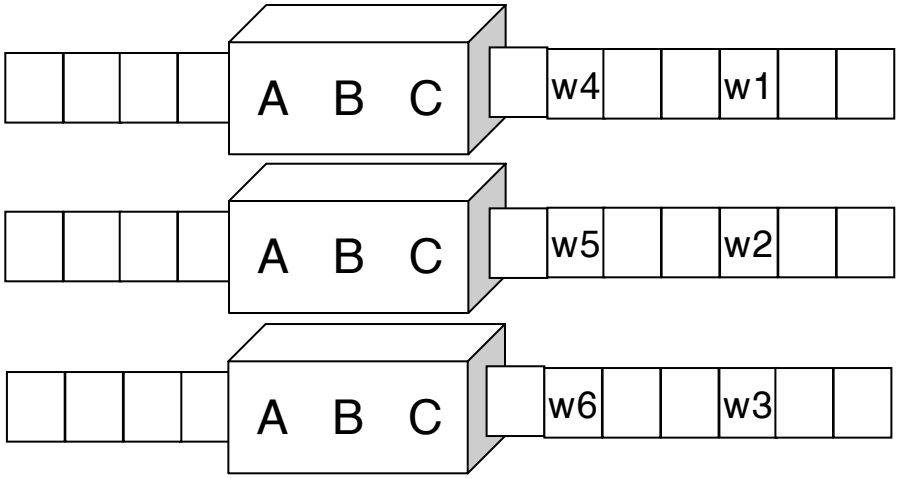
1. Sequential execution



2. Pipeline



3. Data Parallelism



Pipeline & Data Parallelism

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

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

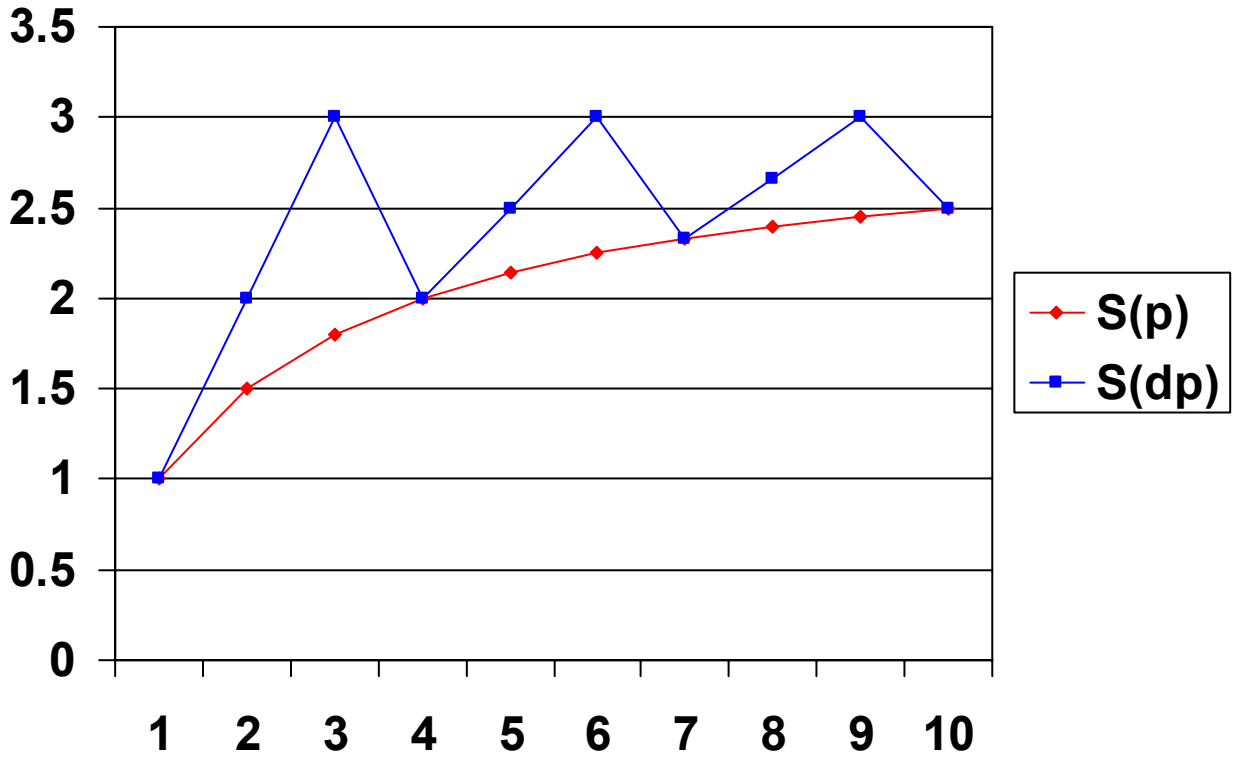
  T(dp): Data-parallelism execution time (with 3 processors)

  S(p): Speedup of pipeline

  S(dp): Speedup of data parallelism

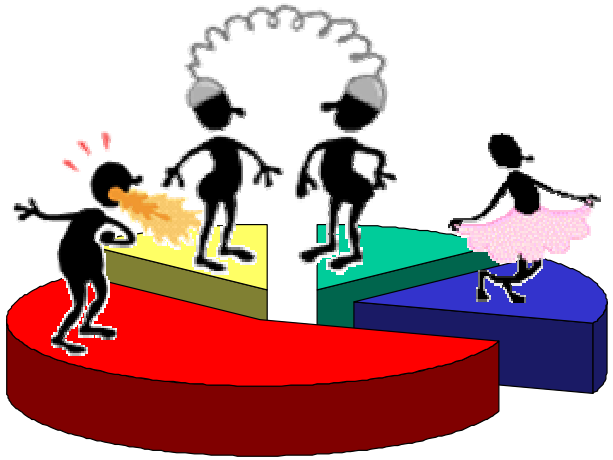
widget	1	2	3	4	5	6	7	8	9	10
T(s)	3	6	9	12	15	18	21	24	27	30
T(p)	3	4	5	6	7	8	9	10	11	12
T(dp)	3	3	3	6	6	6	9	9	9	12
S(p)	1	1+1/2	1+4/5	2	2+1/7	2+1/4	2+1/3	2+2/5	2+5/11	2+1/2
S(dp)	1	2	3	2	2+1/2	3	2+1/3	2+2/3	3	2+1/2

Pipeline & Data Parallelism



Control Parallelism

- Applying different operations to different data elements simultaneously



Scalability

- 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 control-parallelism algorithms