

Programmazione concorrente

Laurea Magistrale in Ingegneria Informatica

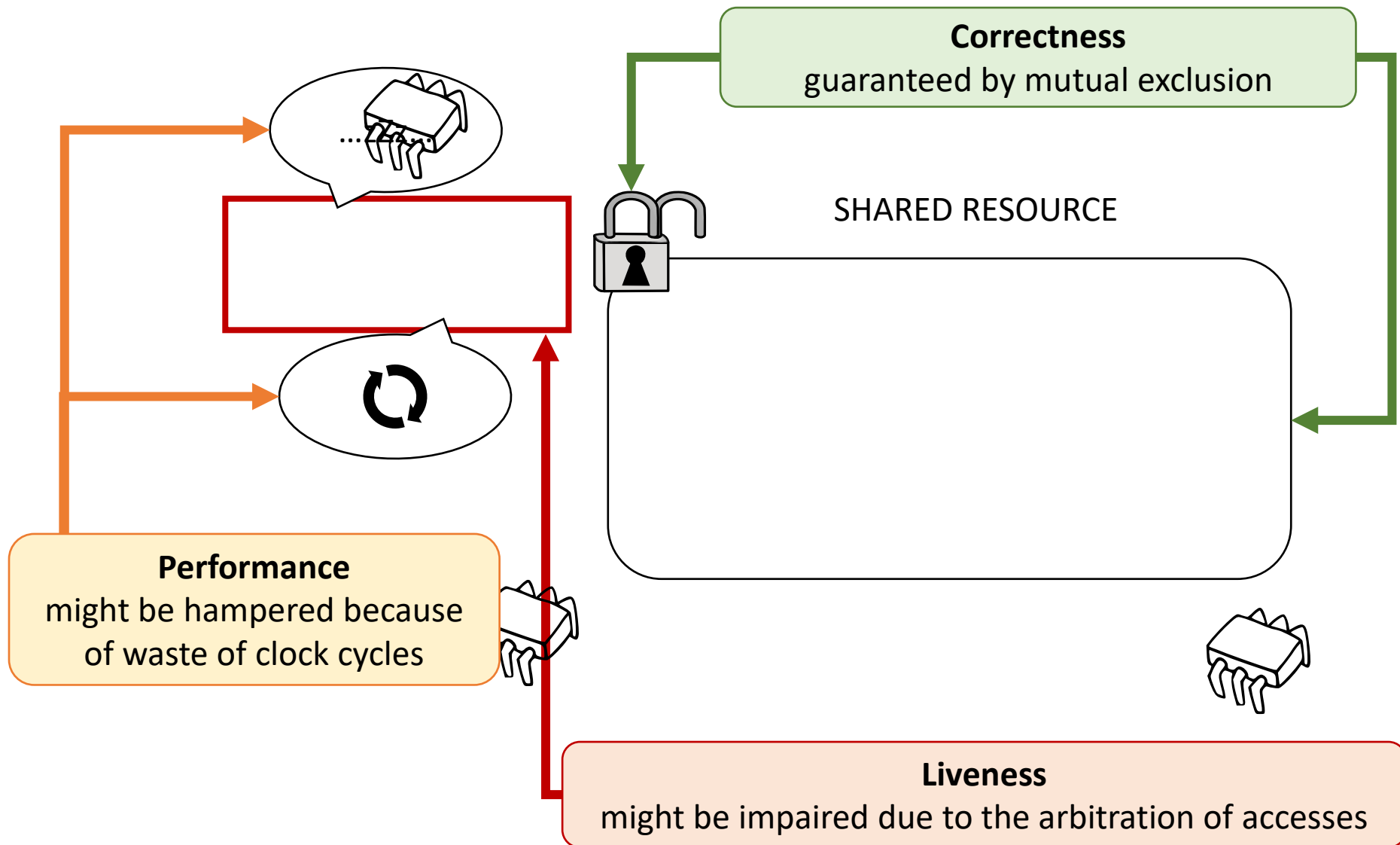
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Properties of Concurrent Programs

1. Speed up
2. Correctness
3. Progress

On concurrent programming



What do we want from parallel programs?

- **Safety:** *nothing wrong happens*
(Correctness)
 - parallel versions of our programs should be correct as their sequential implementations
- **Liveliness:** *something good happens eventually* (Progress)
 - if a sequential program terminates with a given input, we want that its parallel alternative also completes with the same input
- **Performance**
 - we want to exploit our parallel hardware

A bit of terminology

- Hardware
 - Processor
 - CPU
 - CPU-Core
 - Logical Core
 - Hardware thread
- Software
 - Process
 - Thread
 - Fiber
 - Task
- Programs
 - Sequential
 - Concurrent
 - Parallel
 - Distrubuted
- Memory
 - Shared
 - Distributed

The system model

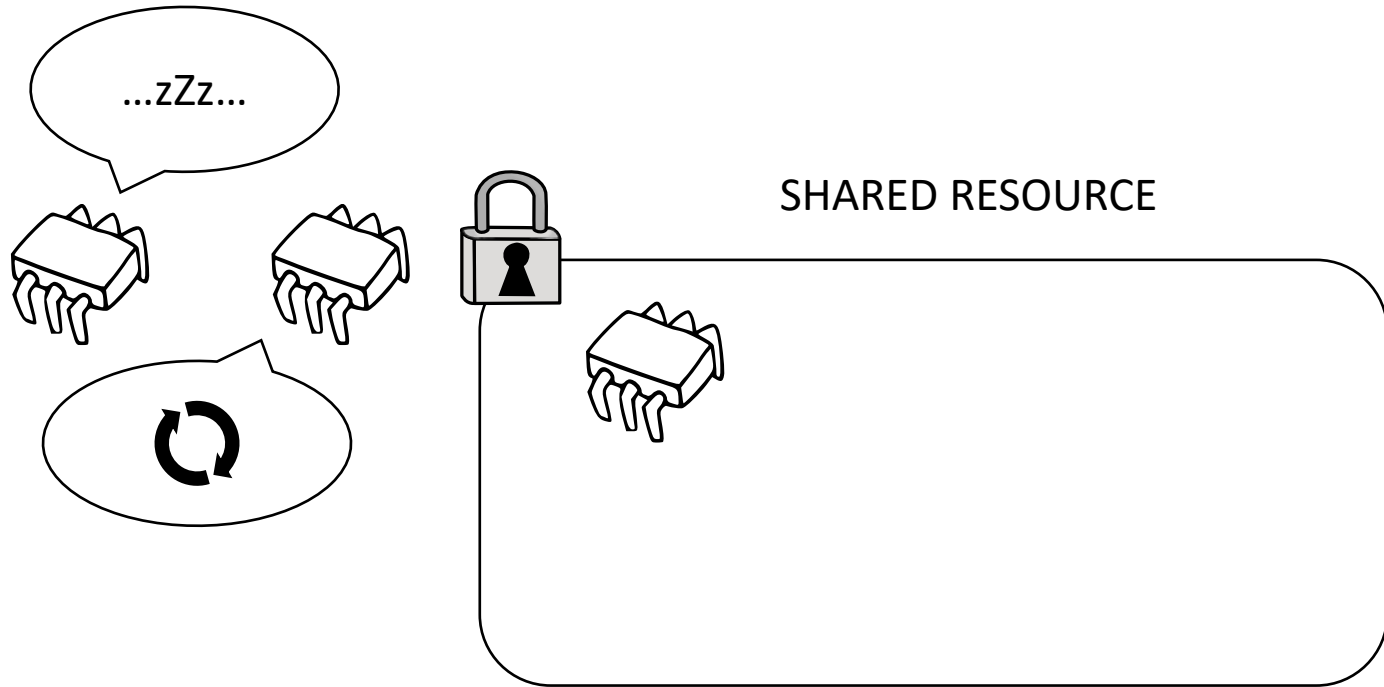
- Threads (aka processes)
- Cores (aka cpus)
- Shared memory
- Arbitrary long asynchronous delays
- Scheduler
 - A system component that decides which/when a thread runs on a given core

Speed up

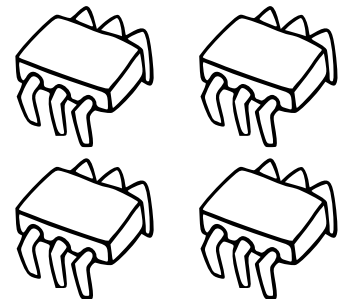
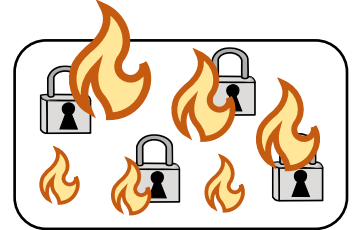
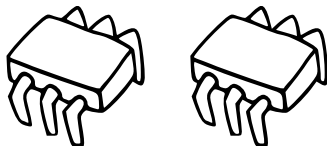
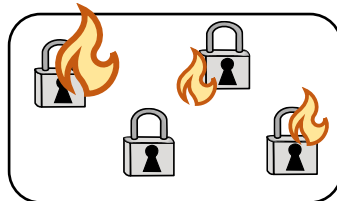
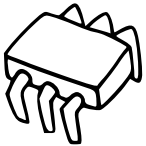
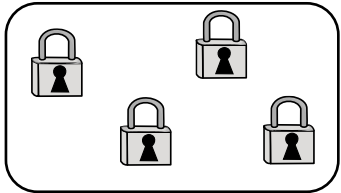
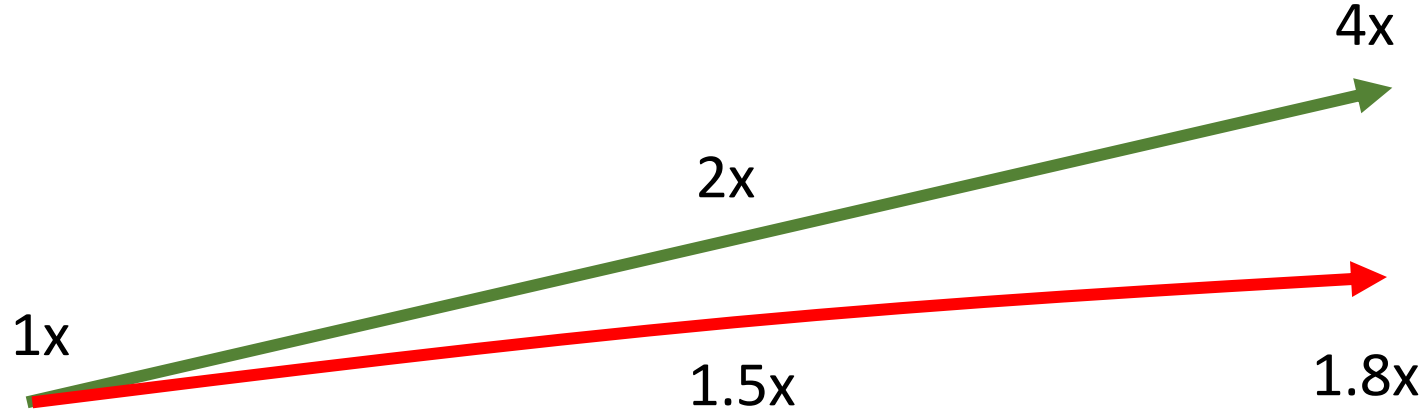
Correctness conditions

Progress conditions

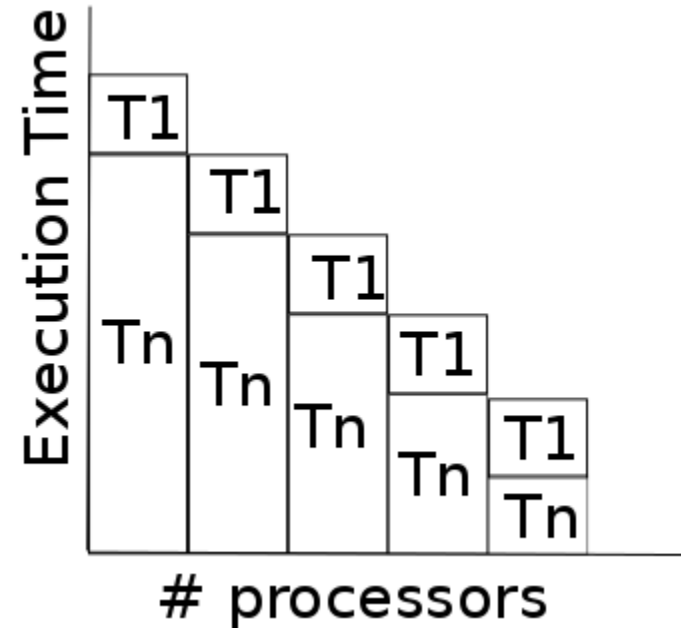
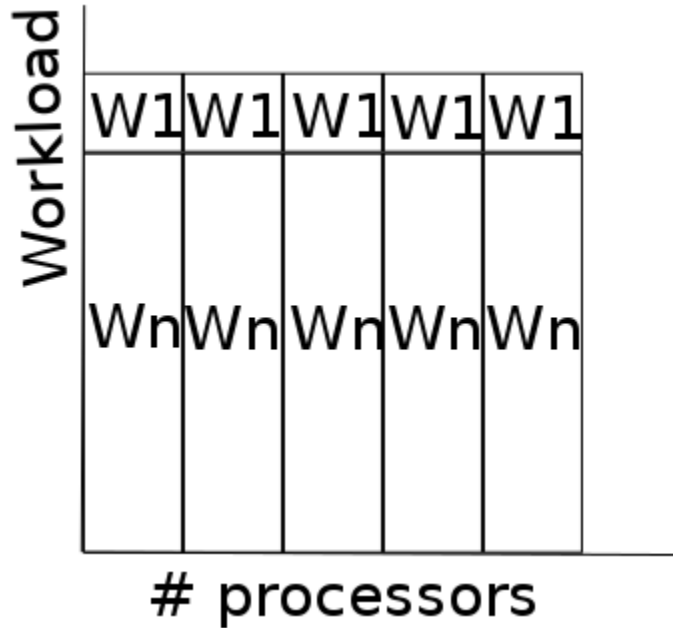
The cost of synchronization



The cost of synchronization



Amdahl Law – Fixed-size Model (1967)



Amdahl Law – Fixed-size Model (1967)

- The workload is fixed: it studies how the behavior of the same program varies when adding more computing power

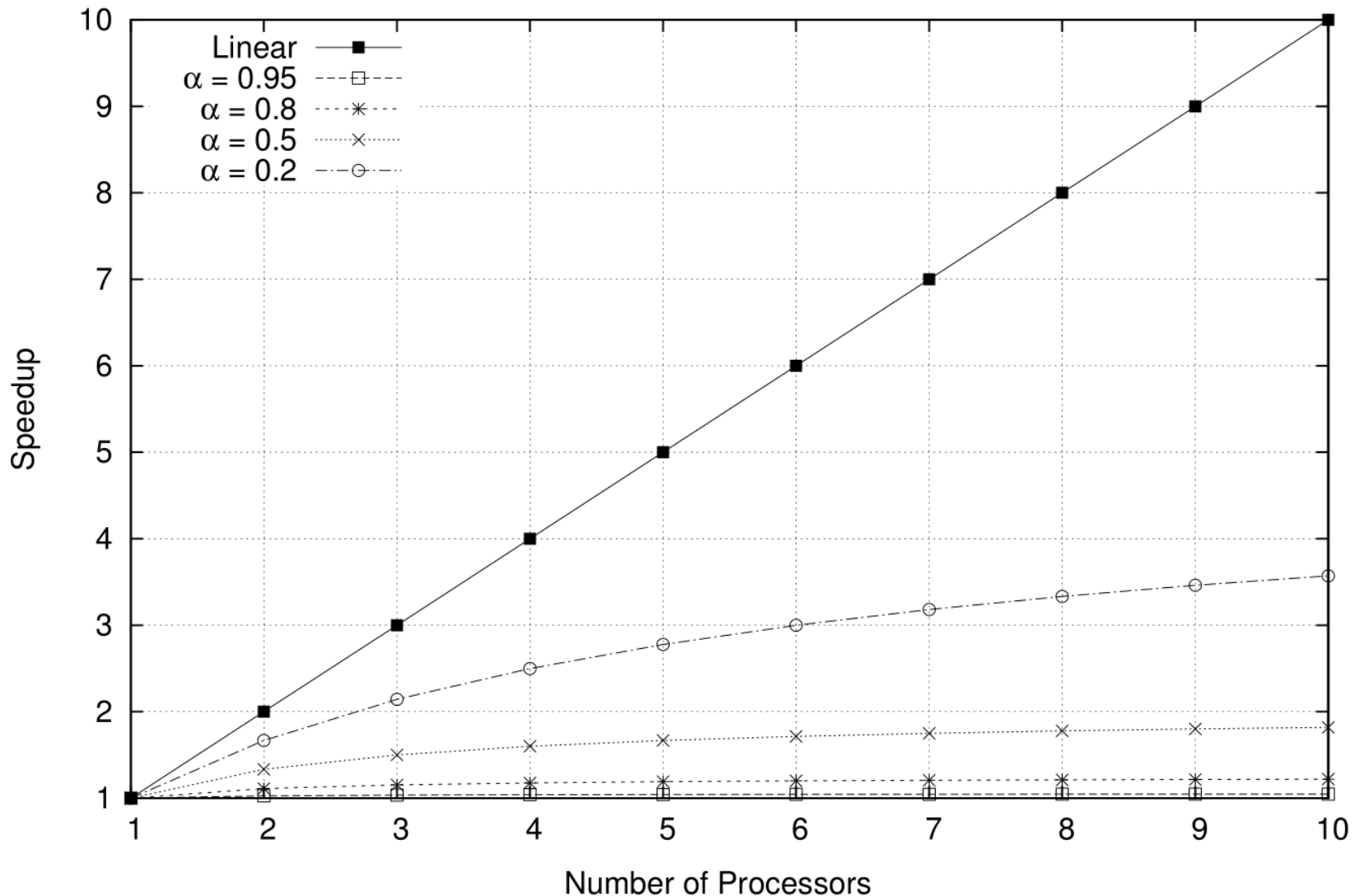
$$S_{Amdahl} = \frac{T_s}{T_p} = \frac{T_s}{\alpha T_s + (1 - \alpha) \frac{T_s}{p}} = \frac{1}{\alpha + \frac{(1 - \alpha)}{p}}$$

- where:
 - $\alpha \in [0,1]$: Serial fraction of the program
 - $p \in N$: Number of processors
 - T_s : Serial execution time
 - T_p : Parallel execution time
- It can be expressed as well vs. the parallel fraction

$$P = 1 - \alpha$$

Amdahl Law – Fixed-size Model (1967)

Parallel Speedup vs. Serial Fraction



How real is this?

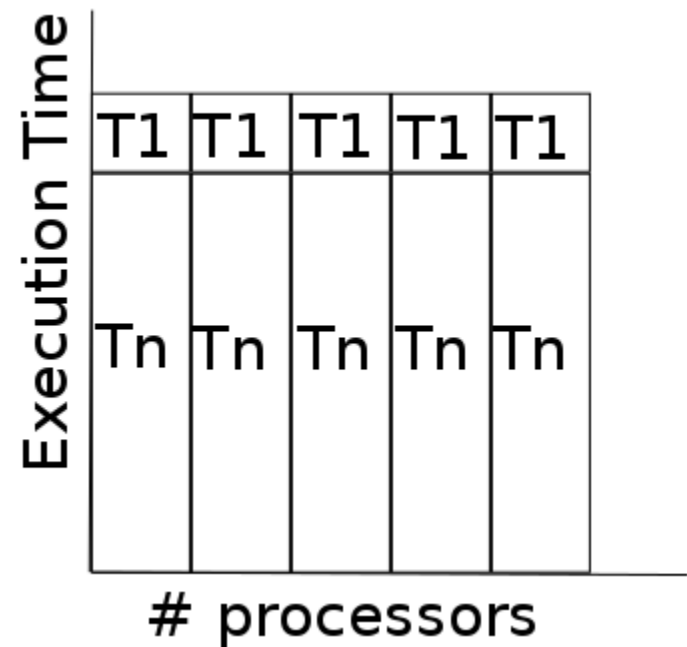
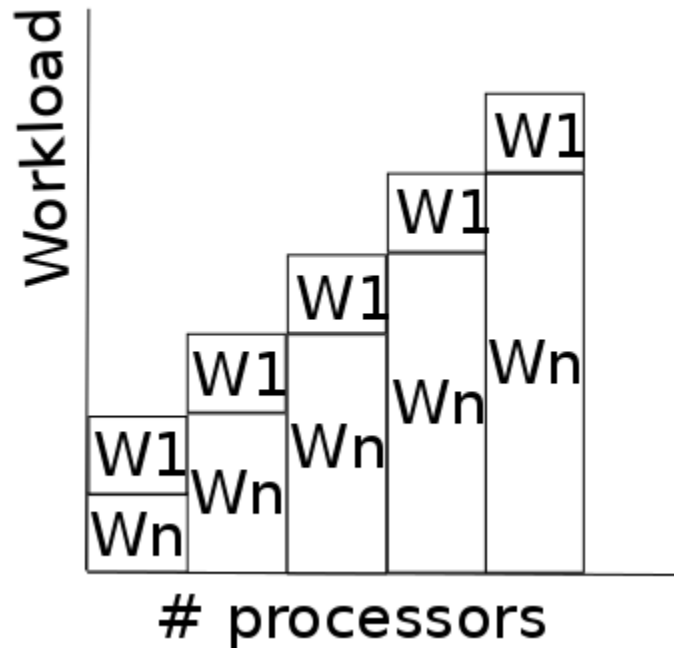
$$\lim_{p \rightarrow \infty} S_{Amdahl} = \lim_{p \rightarrow \infty} \frac{1}{\alpha + \frac{(1 - \alpha)}{p}} = \frac{1}{\alpha}$$

- If the sequential fraction is 20%, we have:

$$\lim_{p \rightarrow \infty} S_{Amdahl} = \frac{1}{0.2} = 5$$

- Speedup 5 using infinite processors!

Fixed-time model



Gustafson Law—Fixed-time Model (1989)

- The execution time is fixed: it studies how the behavior of the scaled program varies when adding more computing power

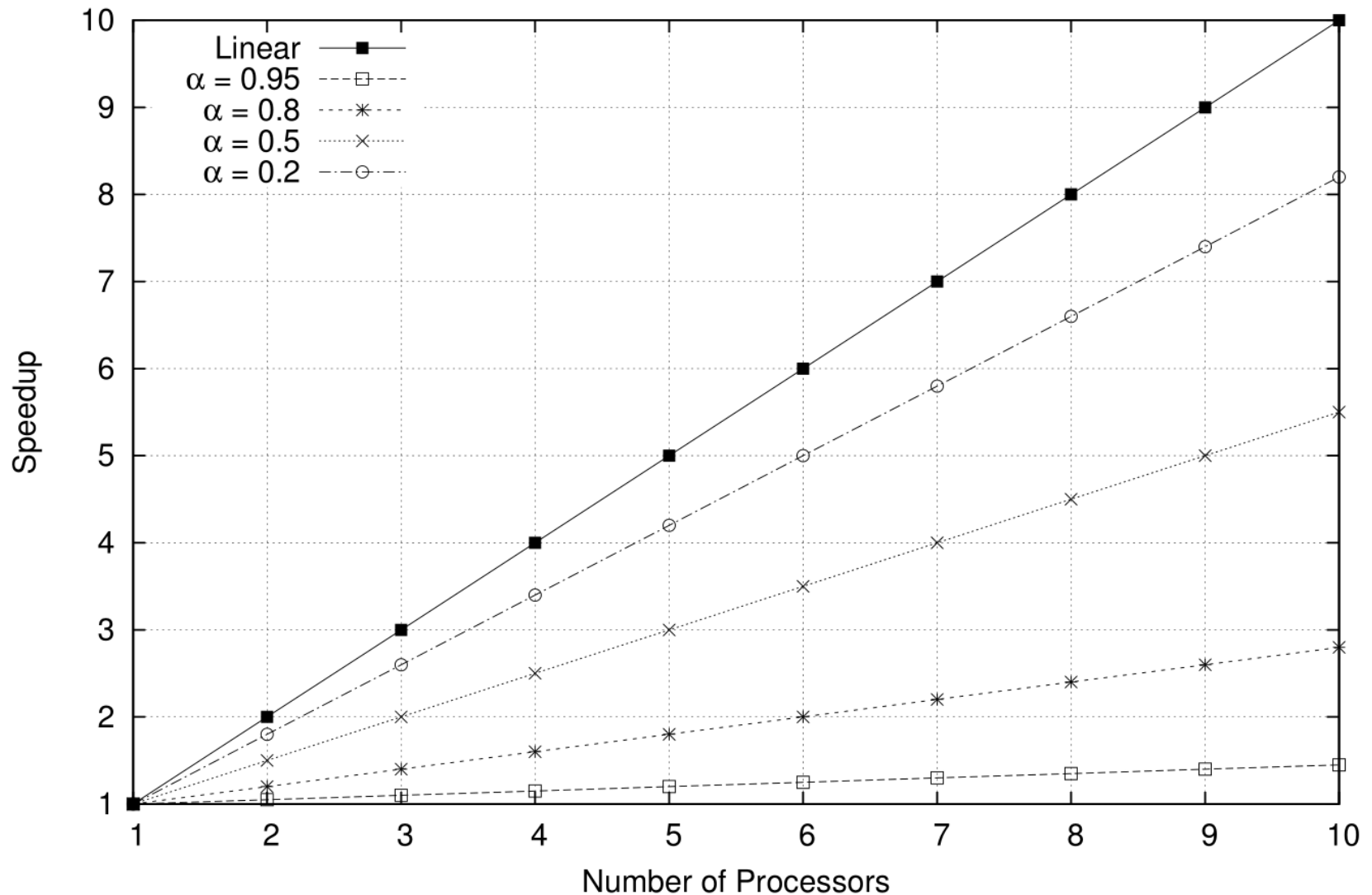
$$W' = \alpha W + (1 - \alpha)pW$$

$$S_{Gustafson} = \frac{W'}{W} = \alpha + (1 - \alpha)p$$

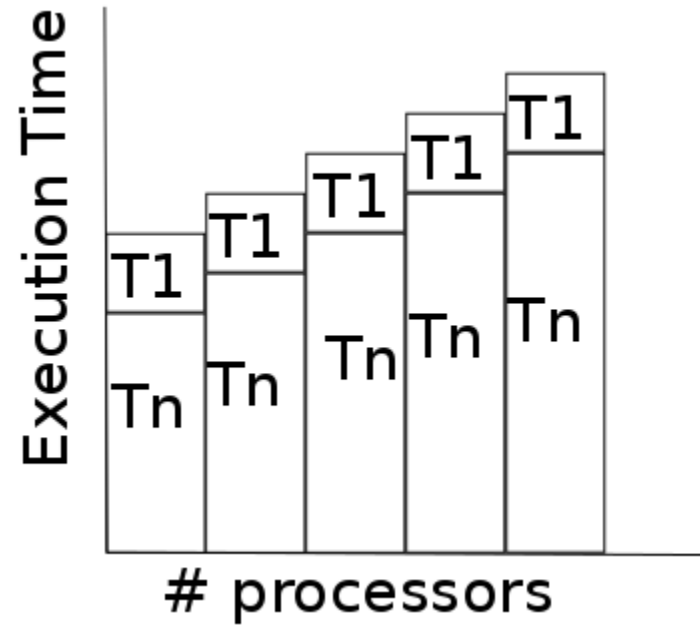
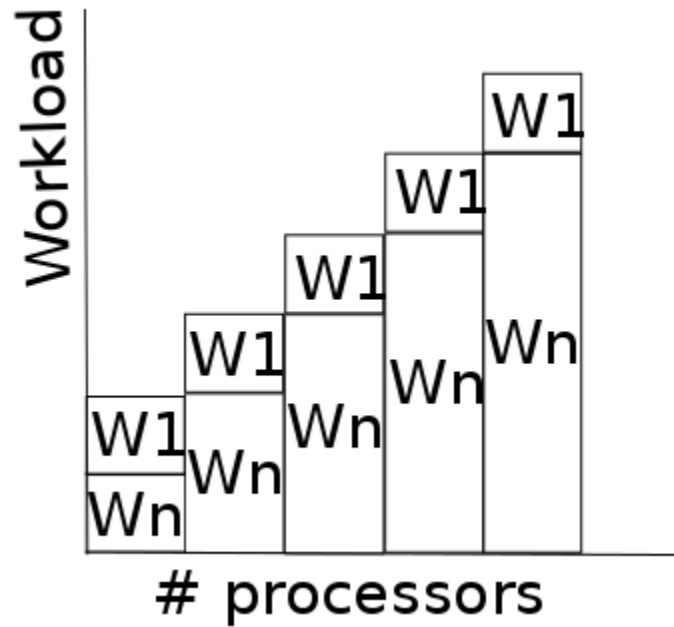
- where:
 - $\alpha \in [0,1]$: Serial fraction of the program
 - $p \in N$: Number of processors
 - W : Original workload
 - W' : Scaled workload

Speed-up according to Gustafson

Parallel Speedup vs. Serial Fraction



Memory-bounded model



Sun Ni Law—Memory-bounded Model (1993)

- The workload is scaled, bounded by memory

$$S_{Sun-Ni} = \frac{\text{sequential time for } W^*}{\text{parallel time for } W^*}$$

$$S_{Sun-Ni} = \frac{\alpha W + (1 - \alpha)G(p)W}{\alpha W + (1 - \alpha)G(p)\frac{W}{p}} = \frac{\alpha + (1 - \alpha)G(p)}{\alpha + (1 - \alpha)\frac{G(p)}{p}}$$

- where:
 - $G(p)$ describes the workload increase as the memory capacity increases
 - $W^* = \alpha W + (1 - \alpha)G(p)W$

Speed-up according to Sun Ni

$$S_{Sun-Ni} = \frac{\alpha + (1 - \alpha)G(p)}{\alpha + (1 - \alpha)\frac{G(p)}{p}}$$

- If $G(p) = 1$

$$S_{Amdahl} = \frac{1}{\alpha + \frac{(1 - \alpha)}{p}}$$

- If $G(p) = p$

$$S_{Gustafson} = \alpha + (1 - \alpha)p$$

- In general, $G(p) > p$ gives a higher scale-up

Superlinear speedup

- Can we have a Speed-up $> p$?
 - Workload increases more than computing power ($G(p) > p$)
 - Cache effect: larger accumulated cache size. More or even all of the working set can fit into caches and the memory access time reduces dramatically
 - RAM effect: enables the dataset to move from disk into RAM drastically reducing the time required, e.g., to search it.

Scalability

- Efficiency

$$E = \frac{\textit{speedup}}{\textit{\#processors}}$$

- **Strong Scalability:** If the efficiency is kept fixed while increasing the number of processes and maintain fixed the problem size
- **Weak Scalability:** If the efficiency is kept fixed while increasing at the same rate the problem size and the number of processes