# Complexity Measures for Parallel Computation

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### Problem parameters:

- n index of problem size
- p number of processors

# Algorithm parameters:

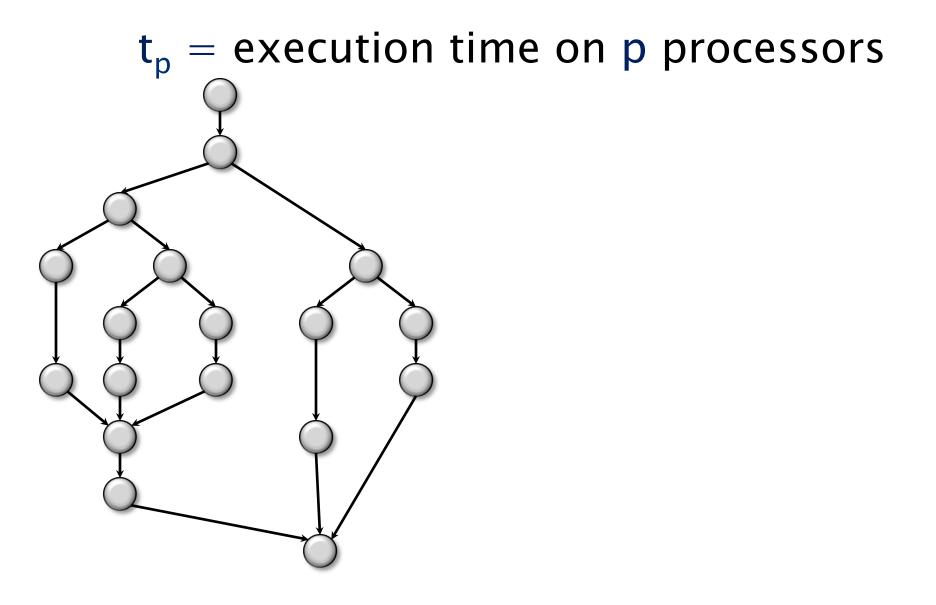
- t<sub>p</sub> running time on p processors
- t<sub>1</sub> time on 1 processor = sequential time = "work"
- $t_{\infty}$  time on unlimited procs = critical path length = "span"
- v total communication volume

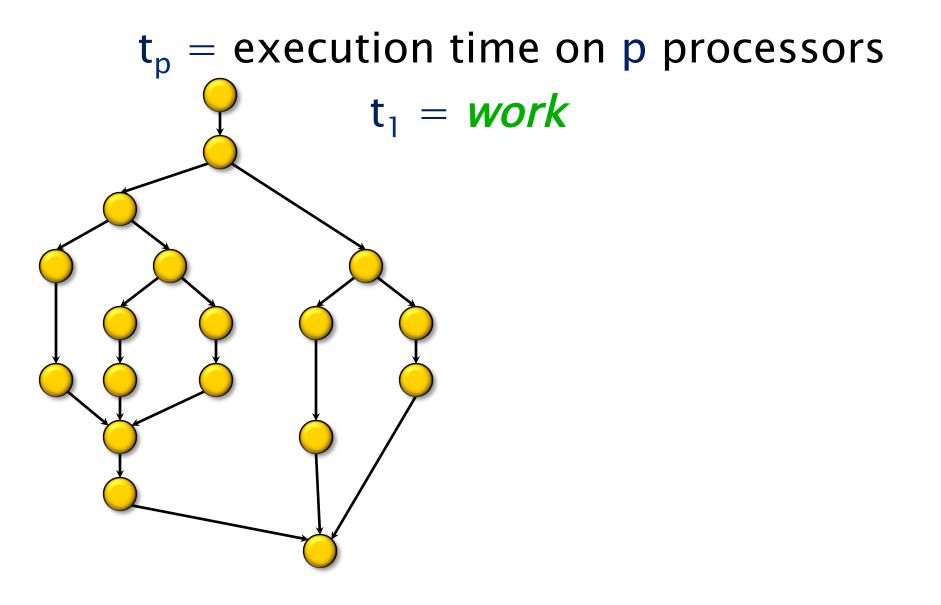
### Performance measures

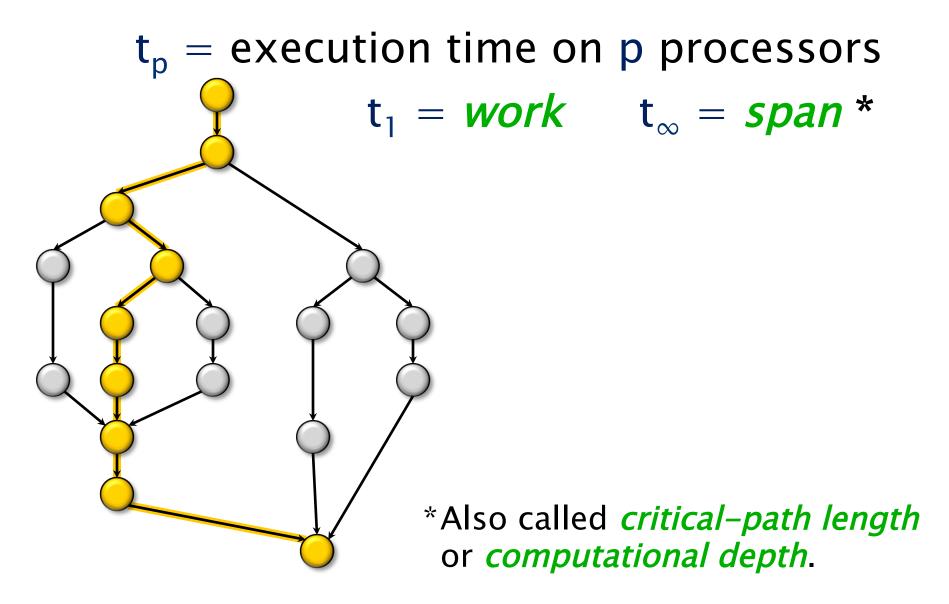
- speedup  $s = t_1 / t_p$
- efficiency  $e = t_1 / (p^*t_p) = s / p$
- (potential) parallelism  $pp = t_1 / t_{\infty}$
- computational intensity  $q = t_1 / v$

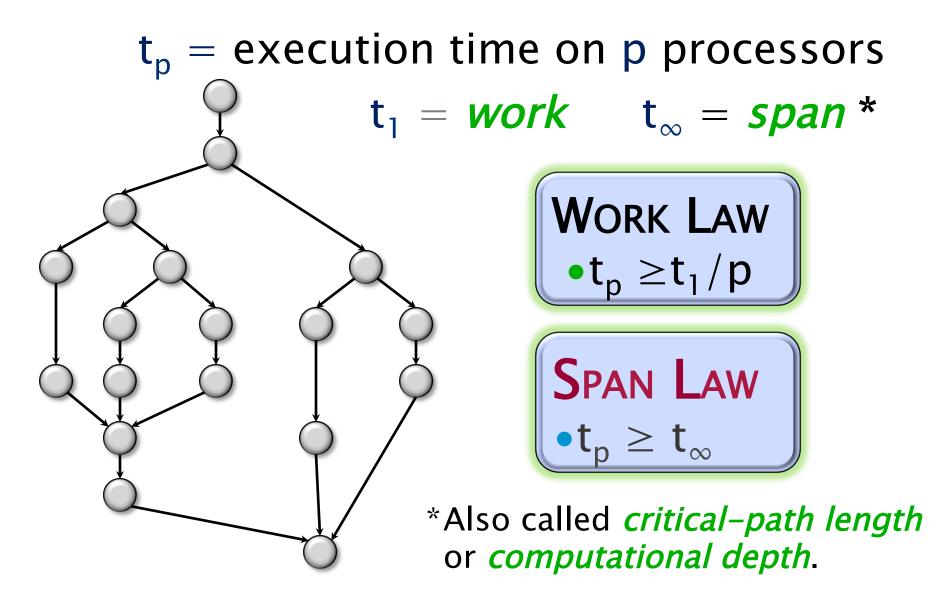
# Several possible models!

- Execution time and parallelism:
  - Work / Span Model
- Total <u>cost</u> of moving data:
  - Communication Volume Model
- Detailed models that try to capture <u>time</u> for moving data:
  - Latency / Bandwidth Model (for message-passing)
  - Cache Memory Model
    (for hierarchical memory)
  - Other detailed models we won't discuss: LogP, UMH, ....

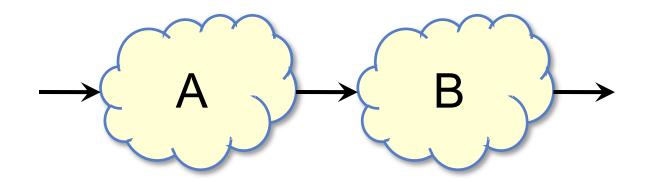






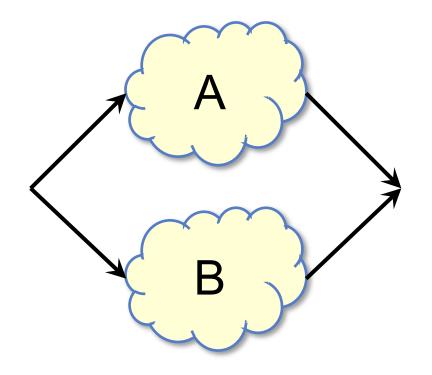


### **Series Composition**



*Work:*  $t_1(A \cup B) = t_1(A) + t_1(B)$ *Span:*  $t_{\infty}(A \cup B) = t_{\infty}(A) + t_{\infty}(B)$ 

### **Parallel Composition**



Work:  $t_1(A \cup B) = t_1(A) + t_1(B)$ Span:  $t_{\infty}(A \cup B) = max\{t_{\infty}(A), t_{\infty}(B)\}$ 



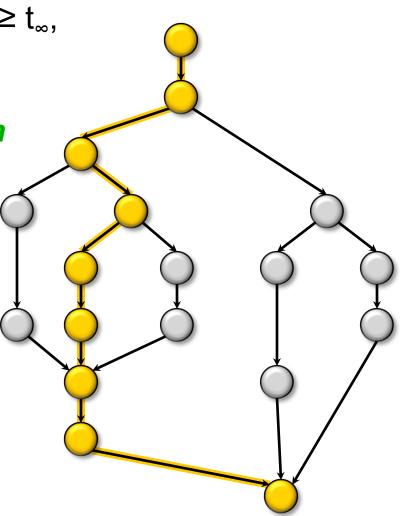
# **Def.** $t_1/t_P = speedup$ on p processors.

If  $t_1/t_P = \Theta(p)$ , we have *linear speedup*, = p, we have *perfect linear speedup*, > p, we have *superlinear speedup*, (which is not possible in this model, because of the Work Law  $t_p \ge t_1/p$ )

### Parallelism

Because the Span Law requires  $t_p \ge t_{\infty}$ , the maximum possible speedup is

- $t_1/t_{\infty}$  = (potential) parallelism
  - the average amount of work
     per step along the span.



# Laws of Parallel Complexity

- Work law:  $t_p \ge t_1 / p$
- Span law:  $t_p \ge t_{\infty}$
- <u>Amdahl's law</u>:
  - If a fraction f, between 0 and 1, of the work must be done sequentially, then

speedup  $\leq 1/f$ 

• Exercise: prove Amdahl's law from the span law.

# **Communication Volume Model**

- Network of p processors
  - Each with local memory
  - Message-passing
- Communication volume (v)
  - Total size (words) of all messages passed during computation
  - Broadcasting one word costs volume p (actually, p-1)
- No explicit accounting for communication time
  - Thus, can't really model parallel efficiency or speedup; for that, we'd use the latency-bandwidth model (see later slide)

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### Performance measures

- speedup  $s = t_1 / t_p$
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- (potential) parallelism  $pp = t_1 / t_{\infty}$
- computational intensity  $q = t_1 / v$

# Detailed complexity measures for data movement I: Latency/Bandwith Model

#### Moving data between processors by message-passing

- Machine parameters:
  - $\alpha$  or  $t_{startup}$  latency (message startup time in seconds)
  - $\beta$  or  $t_{data}$  inverse bandwidth (in seconds per word)
  - between nodes of Triton,  $\alpha \sim 2.2 \times 10^{-6}$  and  $\beta \sim 6.4 \times 10^{-9}$
- Time to send & recv or bcast a message of w words:  $\alpha + w^*\beta$
- t<sub>comm</sub> total communication time
- t<sub>comp</sub> total computation time
- Total parallel time:  $t_p = t_{comp} + t_{comm}$

# Detailed complexity measures for data movement II: Cache Memory Model

Moving data between cache and memory on one processor:

- Assume just two levels in memory hierarchy, fast and slow
- All data initially in slow memory
  - m = number of memory elements (words) moved between fast and slow memory
  - t<sub>m</sub> = time per slow memory operation
  - **f** = number of arithmetic operations
  - $t_f = time per arithmetic operation, t_f << t_m$
  - q = f / m (computational intensity) flops per slow element access
- Minimum possible time =  $f * t_f$  when all data in fast memory
- Actual time

•  $f * t_f + m * t_m = f * t_f * (1 + t_m/t_f * 1/q)$ 

• Larger q means time closer to minimum  $f * t_f$