# Introduction

# **Background and Motivation**

- Parallel computing is a part of HPC.
  - HPC also includes everything else that makes the computation fast.
  - No point parallelizing without increasing performance.
  - You might want to optimize for the architecture.
  - Sometimes overhead outweighs benefits from parallelization.
- Focusing on parallel algorithms.
  - Different version of parallel algorithms suits different architecture or models.
- Many application yo.
- People made super computers throughout the 1900s
- Super computers rely on carefully designed interconnects.
- Cloud computers are just AWS instances.
- Many aspects

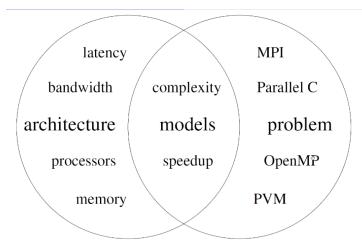


Figure: Overlapping aspects of parallel computing

## Complexity

- $f(n) = O(g(n)) \Rightarrow f$  grows no faster than g
- $f(n) = \Omega(g(n)) \Rightarrow f$  grows no slower than g
- $f(n) = o(g(n)) \Rightarrow f$  grows slower than g
- $f(n) = \omega(g(n)) \Rightarrow f$  grows faster than g
- $f(n) = \Omega(g(n)) \land f(n) = O(g(n)) \Rightarrow f(n) = \Theta(g(n))$
- Strictly speaking we should really use  $\in$  instead of =
- Some common name for complexities:
  - Constant
  - Logarithmic
  - Polylog:  $(\log(n))^c$
  - Linearithmic:  $n \log n$
  - Quadratic:  $n^2$
  - Polynomial or geometric
  - Exponential
  - Factorial
- Log factor are often ignored.

#### Model

- RAM model: random access machine
  - Common model when we talk about sequential time complexity.
- Multiplying the number of computers by a constant factor doesn't change the complexity.
  - Solution: allow p, the number of processors to increase with problem size and hence reduces the complexity.

#### **PRAM**

- Parallel Random Access Machine
- *p* number of RAM processors, each have private memory and share a large shared memory, all memory access takes the same amount of time.
- Does things synchronously, AKA in lock steps.
- PRAM pseudo code looks like regular pseudo code but there's this

$$\label{eq:constraints} \begin{split} & \textbf{for } i \leftarrow 0 \textbf{ to } n-1 \textbf{ do in parallel} \\ & \textbf{processor } i \textbf{ does } thingy \end{split}$$

Many different PRAM model

- EREW: exclusive read, exclusive write
- CREW: concurrent read, exclusive write
- CRCW: concurrent read, concurrent write
  - Concurrent write have different types
    - COMMON: Error when two processor tries to write to the same location with different value.
    - ARBITRARY: Pick a arbitrary processor if many processor writes the same time.
    - PRIORITY: Processor with lowest ID writes.
    - COMBINING: Runs a function whenever multiple processors tries to write at the same time.
      - Too powerful.
- ERCW: exclusive read, concurrent write (never used)

Power of model: expresses the set of all problems that can be solved within a certain complexity.

- A is more powerful that B if A can solve a larger set of problems within any complexities.
- A is equally powerful as B if they can solve the same set problems within any complexities.
- Partial ordering.
- COMMON, ARBITRARY, PRIORITY and COMBINING are in increasing order of power.
- Any CRCW PRIORITY PRAM can be simulated by a EREW PRAM with a complexity increase of  $\mathcal{O}(\log p)$
- *Parallel Computation Thesis*: any thing can be solved with a Turing Machine with polynomially bounded space can be solved in polynomially bounded space with unlimited processors.
  - Unbounded *word sizes* are not useful, so we limit word counts to  $\mathcal{O}(\log p)$
- *Nick's Class* (NC): Solvable in polylog time with ploy number of processors.
- Widely believed that  $\mathbf{NP} \neq P$

## Definitions (need to remember)

- $w(n) = t(n) \times p(n)$  where w(n) is the work / cost, t(n) is the time and p(n) is the number of processors.
  - Optimal processor allocation means:  $t(n) \times p(n) = \Theta(T(n))$  where T(n) is the time taking by a sequential algorithm.
    - Equivalent to  $t(n) \times p(n) = O(T(n))$  because  $t(n) \times p(n) = \Omega(T(n))$  always.
  - Speedup $(n) = \frac{T(n)}{t(n)}$ 
    - Speedup optimal = processor optimal.
  - Optimal: processor optimal AND  $t(n) = \mathcal{O}(\log^k n)$ 
    - Processor optimal and polylog in time.
  - Efficient: Assume  $T(n) = \Omega(n) w(n) = \mathcal{O}(T(n) \log^{\alpha} n)$  AND polylog in time
    - Optimal but polylog increase in work.
- size: Size(n) is the total number of operations it does.
- efficiency:  $\eta(n)$  speedup per processor  $\eta(n) = \frac{T(n)}{w(n)} = \frac{\operatorname{Speedup}(n)}{p(n)}$
- You can decrease p and increase t by a factor of  $O\left(\frac{p_1}{p_2}\right)$ , w(n) doesn't increase its complexity.
  - Can't do it the other way around.

## **Brent's Principle (important)**

• If something can be done with size x and t time with infinite processors, then it can be done in  $t + \frac{x-t}{n}$  time with p processors

#### Amdahl's Law

- Maximum speedup: if f is the fraction of time that can't be parallelized, then Speedup $(p) \to \frac{1}{f}$  as  $p \to \infty$ 
  - Honestly very obvious.

### Gustafson's Law

- s is fraction time of serial part, r is fraction time of parallel part, then Speedup $(p) = \Omega(p)$ 
  - Very obvious again...

### Algorithms

- sum
- · logical or
- Maximum
  - $n^2$  processors all compare all elements and set is max array to false if element isn't maximum.
  - Only processor with element being max write it to the returning memory address.
- Maximum $n^2$ 
  - $\mathcal{O}(\log \log n)$
  - n processor on n elements.
  - Is efficient
  - Make elements into a square, find maximum on each row recursively.
  - Find maximum of maximum of the rows using maximum.
  - $\mathcal{O}(\log \log n)$  levels of recursion, each level takes  $\mathcal{O}(1)$  times
- Element Uniqueness
  - Have an array size of MAX INT.
  - Write processor ID to the array with the element.
  - Check if processor ID is indeed there, if not there's another element there.

- Replication
  - $O(\log n)$
- Replication optimal
- $p=\frac{n}{\log(n)}$  and copy at the end. Broadcast
- - Just replicate
- Simulate PRIORITY with COMMON  $n^2$ 
  - Minimum version of Maximum
- Simulate PRIORITY with EREW
  - All processor wants to write
  - Sort array A of tuples (address, processorID) using Cole's Merge Sort.
  - For each processor k, if A[k].address  $\neq A[k-1]$ .address then A[k].processor D is the smallest ID that wants to write to that address.