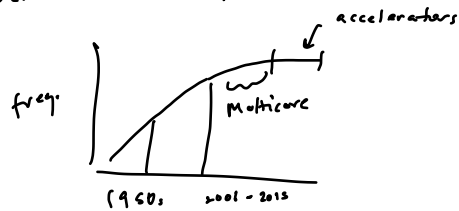


"How to have a bad research career"

- 3 homeworks - 30%
 - project - 40%
 - midterm - 25%
 - 5% participation
-

Why study parallel algorithms and d.s.?

- Moore's law ending " Dennard scaling "



- parallelism & concurrency fundamental ideas for how to do work (e.g. producing a complex object)
- superset of serial algorithms; understand new aspects of a problem by studying in parallel.

- Amount of data to analyze not slowing down.
 - genomics data
 - video/image data

(i) Multicore machines exist!

- p cores, $p \times$ faster
factor p speedup.

$$\frac{t_1}{t_p} \quad \left(\frac{t^*}{t_p} \right) \leftarrow \text{best } \underline{\text{seq.}} \text{ running time}$$

$O(n^2)$ operations
(work)

$O(n \log n)$ operations

- Early apps were large-scale simulation

1961

IBM Stretch:

1MB memory

}

1972

Illiac IV:

4 CPUs, 256 FPU's

512K memory

1 CPU 64 FPU's

→ SIMD (single-instruction, multiple-data)

1973

Eispach: matrix eigenvalues

↳ Lapack (Jack Dongarra)

1976

CRAY-1

- vector processing capabilities

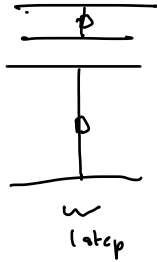
□ □ ⊕ ⊗

1912 :

MSF algorithm (Binckle)

1968 :

Batcher Sort / Bitonic Sort



$O(n \log^2 n)$ operations

$O(\log^2 n)$ delay (# steps)
rounds

1975 :

Valiant: Max, Merging and Sorting

• $O(\log \log n)$ rounds $\underbrace{\Omega(\log \log n)}_{n \text{ processors}}$ round

$\Omega(\log n)$

n processors

1978 :

PRAM (wylie)

1979 : Circuit models (P-completeness, NC)

1980s: Golden age.

90-95: other model (Asynch PRAM)

(BSP - Valiant)

↑

predecessor of MapReduce

1995 Work/depth or Work/Span model.

}
}

Parallelism winter

2010

}
}

Resurgence - Lot of interesting work.

Present

RAM: Random Access Machine;

- how does an algorithm behave when data grows?

$O(n^2)$ time vs $O(n \log n)$ time.

- +, x, ÷ unit time

→ $O(1)$

- load/store unit time

- load: compare algorithms asymptotically

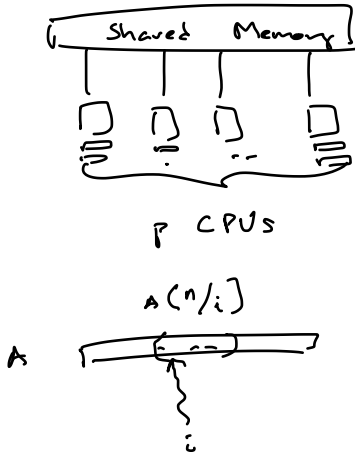
- $O(\log n)$ bit words

┌ Main Memory ─┐



Parallel Models

PRAM | Parallel Random Access Machine



- synchronous
- works in lockstep
- not particularly realistic
 - fixed # processors
 - schedule tasks \rightarrow processors?

$$\text{Cost} = \frac{T}{P} = \frac{\text{Running Time}}{\text{Processors}} \quad (p \text{ processors})$$

$$\text{total \# instructions} = PT = \text{work}$$

$$\text{Sorting in } O\left(\frac{n \log n}{P} + \log^2 n\right) \text{ time}$$

PRAM models: ER : can't read same location at the same time

EW : " write "

CR

CW

EREW, CREW, CRCW

\downarrow

Common

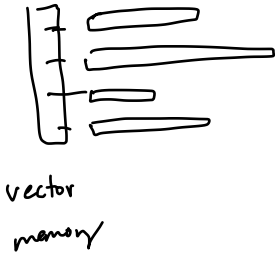
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Priority

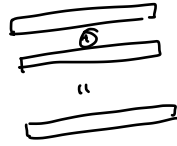
Vector Models

Connection Machine

Guy Blelloch's thesis



- instructions operate on vectors



merge,
prefix-sum

Cost: Work (element complexity) = $\sum_{s_i \in \text{Steps}} \text{length}(v_i)$

Step complexity = # steps executed by the program

Eg. sorting in $O(\log^2 n)$ steps
 $O(n \log n)$ work

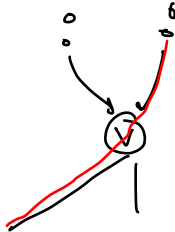
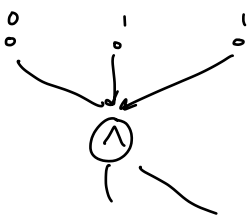
"Connection Machine" Book

NESL

Circuit Model 1979 (Pippenger)

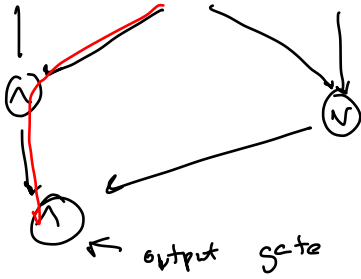
- view parallel comp. as a circuit (DAG)

\wedge (and) and \vee (or) gates



Input

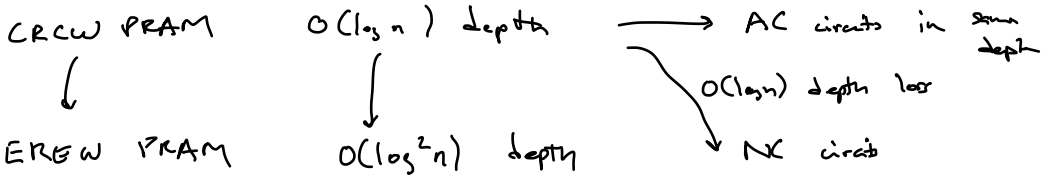
$\in \{0, 1\}$



Cost : - size = # gates
 - depth

Sorting : (AKS network)

$O(n \log n)$ size, $O(\log n)$ depth



NC^k : polynomial size circuits with $O(\log^k n)$ depth

$NC = \bigcup_k NC^k$