

Complexity Theory

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Based on slides by Jörg Kreiker

Lecture 23

NC and AC scrutinized

Recap

Efficient parallel computation

- computable by some PRAM with
- polynomially many processors in
- polylogarithmic time
- robust wrt to underlying PRAM model

Recap

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corresponds to

small depth circuits

- of polynomial size
- polylogarithmic depth
- logspace uniform

Recap – NC and AC

If $L \subseteq \{0, 1\}^*$ is decided by a **logspace-uniform** family $\{C_n\}$ of **polynomially sized** circuits with **bounded fan-in**

- and **depth** $\log^k n$ then $L \in \mathbf{NC}^k$ for $k \geq 0$
- $\mathbf{NC} = \bigcup_{k \geq 0} \mathbf{NC}^k$

Recap – NC and AC

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If the fan-in is **unbounded** we obtain the corresponding **AC** hierarchy.

Goal

Find the places of **NC** and **AC** among other complexity classes!

Agenda

- **NC** versus **AC**
- **NC** versus **P**
- **NC¹** versus **L**
- **NC²** versus **NL**

Unbounded \rightarrow bounded fan-in

Theorem

For all $k \geq 0$

$$\text{NC}^k \subseteq \text{AC}^k \subseteq \text{NC}^{k+1}$$

Unbounded \rightarrow bounded fan-in

Theorem

For all $k \geq 0$

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Proof

- first inclusion trivial
 - for the second, assume $L \in \text{AC}^k$ by family $\{C_n\}$
 - there exists a polynomial $p(n)$ such that
 - C_n has $p(n)$ gates with
 - maximal fan-in of at most $p(n)$
 - each such gate can be simulated by a **binary tree** of gates of the same kind with depth $\log(p(n)) = O(\log n)$
- \Rightarrow the resulting circuit has size at most size $p(n)^2$, depth at most $\log^{k+1} n$ and maximal fan-in 2

Corollary

Theorem

$AC = NC$

Corollary

Theorem

$$AC = NC$$

Remarks

- the inclusions in the theorem on the previous slide are strict for $k = 0$
- one strict inclusion is trivial, the other one is subject of the next lecture
- for practical relevance, we focus on **bounded fan-in**, ie. **NC**

Agenda

- NC versus AC ✓
- NC versus P
- NC¹ versus L
- NC² versus NL

NC versus P

Theorem

NC \subseteq P

Proof

- let $L \in \text{NC}$ by circuit family $\{C_n\}$
- \Rightarrow there exists a **logspace TM** M that computes $M(1^n) = \text{desc}(C_n)$
- the following **P** machine decides L
 - on input $x \in \{0, 1\}^n$ simulate M to obtain $\text{desc}(C_n)$
 - C_n has input variables z_1, \dots, z_n
 - evaluate C_n under the assignment σ that maps z_i to the i -th bit of x
 - output $C_n(\sigma)$
- all steps take **polynomial time** (evaluation takes time proportional to circuit size)

Remarks

- **P** equals the set of languages with **logspace-uniform** circuits of **polynomial size** and **polynomial depth** (exercise)
- it is an **open problem** whether the previous inclusion is strict
- in fact it is open whether **NC¹ ⊂ PH**
- problem is important, since it answers whether **all problems in P** have fast **parallel algorithms**
- conjecture: strict

Agenda

- NC versus AC ✓
- NC versus P ✓
- NC¹ versus L
- NC² versus NL

Proof Steps

1. logspace reductions are **transitive**
2. if $L \in \mathbf{NC}^1$ then there exists a logspace uniform family of circuits $\{C_n\}$ of **depth $\log n$**
3. **circuit evaluation** of a circuit of depth d and **bounded fan-in** can be done in space $O(d)$

What is the theorem?

What is the theorem?

Theorem

$NC^1 \subseteq L$.

Proof

- for a language $L \in NC^1$, we can compute its circuits (step 2) in logspace
- we can evaluate circuits in logspace (step 3)
- the composition of these two algorithms is still logspace (step 1)
- steps 1 and 2 already proven

Proof of Step 3

- evaluate the circuit **recursively**
- identify gates with **paths** from output to input node
 - output node: ϵ
 - **left predecessor** of gate π : $\pi.0$
 - **right predecessor** of gate π : $\pi.1$

Proof of Step 3

- evaluate the circuit **recursively**
- identify gates with **paths** from output to input node
 - output node: ϵ
 - **left predecessor** of gate π : $\pi.0$
 - **right predecessor** of gate π : $\pi.1$
- 1. if π is an **input** return value
 2. if π denotes an **op** gate, compute value of $\pi.0$, value of $\pi.1$ and combine
- recursive depth $\log n$, only one global variable holding current path: total $\log n$ space
- note that the **naive** recursion takes $\log^2 n$ space!

Agenda

- NC versus AC ✓
- NC versus P ✓
- NC¹ versus L ✓
- NC² versus NL

The theorem

Theorem

$$\text{NL} \subseteq \text{NC}^2$$

Proof outline

- show that $\text{Path} \in \text{NC}^2$
- let $L \in \text{NL}$ and NL machine M deciding it; for a given input $x \in \{0, 1\}^*$
- build a circuit C_1 computing the **adjacency matrix** of M 's **configuration graph** on input x
- build a second circuit C_2 that takes this output and decides whether there is an **accepting run**
- the **composition** of C_1 and C_2 decides L
- observe: the composition can be computed in **logspace**

Path \in NC²

- let A be the $n \times n$ adjacency matrix of a graph
- let $B = A + I$ (add self loops)
- compute the square product B^2

$$B_{i,j}^2 = \bigvee_k B_{i,k} \wedge B_{k,j}$$

- contains 1 iff there is a path of length at most 2
 - can be done in $AC^0 \subseteq NC^1$
 - $\log n$ times repeated squaring
- \Rightarrow paths can be computed in NC²

Agenda

- NC versus AC ✓
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- NC¹ versus L ✓
- NC² versus NL ✓

Criticism of NC

The notion of **NC** as **efficient parallel** computation may be criticized.

- **polynomially many** processors
 - in the **NC** hierarchy a $\log n$ algorithm with n^2 processors is favored over one with n processors and time $\log^2 n$
 - expensive
- **polylogarithmic depth**
 - for many **practical** inputs, **sublinear** algorithms might be as good or better
 - e.g. $n^{0.1}$ is at most $\log^2 n$ for values up to 2^{100}

Summary

- $AC = NC$
- $NC^1 \subseteq L \subseteq NL \subseteq NC^2 \subseteq P$
- up next: $AC^0 \subset NC^1$