

Solution

Computational Complexity – Homework 12

Discussed on See website / as otherwise arranged.

Definition 1. A language L is in $\mathbf{P}_{/\text{poly}}$ if there exist a family $\{C_n\}$ of Boolean circuits of size polynomial in n such that for all $x \in \{0, 1\}^n$

$$x \in L \text{ iff } C_n(x) = 1.$$

A family of Boolean circuits $\{C_n \mid n \in \mathbb{N}\}$ is *logspace uniform* if there is a deterministic Turing machine M running in logarithmic space which on input 1^n outputs a description of C_n . Similarly for *polytime uniform* we require M run in polynomial time.

(Note that the definition of \mathbf{NC} requires the logspace uniformity together with polynomial size and polylog depth.)

Exercise 12.1

Show that $\mathbf{BPP} \subseteq \mathbf{P}_{/\text{poly}}$.

Remark: Use one of the results on \mathbf{BPP} which have already been shown in the lecture.

Solution: We have seen in the lecture that if $L \in \mathbf{BPP}$ is decided by some TM $M(x, u)$ then for every $n \in \mathbb{N}$ there exists a $u_n \in \{0, 1\}^{p(n)}$ s.t.

$$\forall x \in \{0, 1\}^n : x \in L \text{ iff } M(x, u_n) = 1.$$

We therefore can first transform $M(x, u)$ into a family of circuits of size polynomial in x (recall that $|u| \leq p(|x|)$) and then hardwire u_n into the circuits.

Exercise 12.2

(a) Show that for every polynomial p the following language is in \mathbf{coNP} :

$$L_p := \left\{ \langle C_1, C_2, \dots, C_n \rangle \mid \begin{array}{l} C_i \text{ is a circuit of size at most } p(i) \text{ which decides SAT for every formula} \\ \text{of length exactly } i \end{array} \right\}.$$

Remark: Assume w.l.o.g. that every formula has length at least one with 0 (false) and 1 (true) the two formulae of length 1. Now, use the circuits C_1, \dots, C_i ($i \geq 0$) to check the correctness of circuit C_{i+1} . (Recall the so-called self-reducibility of SAT.)

- (b) Show that **PH** collapses to the second level if $\mathbf{NP} \subseteq \mathbf{P}/\text{poly}$, i.e. if there is a sequence of polynomial sized circuits for SAT.

Remark: It suffices to show that $\Pi_2\text{SAT} \in \Sigma_2^p$.

- (c) What happens if there is a sequence of polynomial sized circuits for SAT that is moreover logspace uniform?
What if it is polytime uniform?

Solution:

- (a) The TM first checks that $|C_i| \leq p(i)$ and that C_i indeed encodes a Boolean circuit. This takes time $\mathcal{O}(n \cdot p(n))$ (assuming that $p(n)$ grows monotonically). Next, the TM chooses some Boolean formula ϕ of length at most n . If $|\phi| = 1$, the TM directly checks if $C_1(\phi) = \phi$. Otherwise, it choose (deterministically) some variable x from ϕ and checks again in polynomial time that

$$C_{|\phi|}(\phi) = \bigvee_{b=0,1} C_{|\phi[x:=b]|}(\phi[x := b]).$$

By definition of **coNP** the TM accepts $\langle C_1, \dots, C_n \rangle$ only if it does not find any formula of length at most n for which this test fails.

- (b) If $\mathbf{NP} \subseteq \mathbf{P}/\text{poly}$, then there exists some polynomial p and a circuit sequence $(C_n)_{n \in \mathbb{N}}$ with $|C_n| \leq p(n)$ which decides SAT. In particular, for this p every (encoding of a) prefix $\langle C_1, \dots, C_n \rangle$ is in L_p .

Now, $\Pi_2\text{SAT}$, i.e.,

decide if $\forall u \exists v : \phi(u, v)$ is true (with ϕ a Boolean expression)

is Π_2^p -complete where u and v are bounded by the length of ϕ .

We now have

$$\forall u \exists v : \phi(u, v) \text{ iff } \exists \langle C_1, \dots, C_{|\phi|} \rangle \in L_p : \forall u : C_{|\phi(u, \cdot)|}(\phi(u, \cdot))$$

where the latter describes a computation in Σ_2^p . So, $\Pi_2^p \subseteq \Sigma_2^p$ which implies $\mathbf{PH} = \Sigma_2^p \cap \Pi_2^p$.

- (c) In both cases, we have $\mathbf{P} = \mathbf{NP}$. It suffices to show that SAT is then in **P**. Given a formula ϕ of length $n = |\phi|$, we construct in log-space from the input 1^n the circuit C_n which decides SAT for all formulae of length exactly n . As $\mathbf{L} \subseteq \mathbf{P}$, this can be done in time polynomial in $n = |\phi|$. We then evaluate C_n on ϕ . This can again be done in time polynomial in the size of C_n which by definition is polynomial in the size of $n = |\phi|$.

Exercise 12.3

Prove that for $n \geq 100$, most of the boolean functions on n variables require circuits of size at least $2^n/n$.

Solution:

THEOREM 6.15

For $n \geq 100$, almost all boolean functions on n variables require circuits of size at least $2^n / (10n)$.

PROOF: We use a simple counting argument. There are at most s^{3s} circuits of size s (just count the number of labeled directed graphs, where each node has indegree at most 2). Hence this is an upperbound on the number of functions on n variables with circuits of size s . For $s = 2^n / (10n)$, this number is at most $2^{2^n / 10}$, which is miniscule compared 2^{2^n} , the number of boolean functions on n variables. Hence most Boolean functions do not have such small circuits. ■

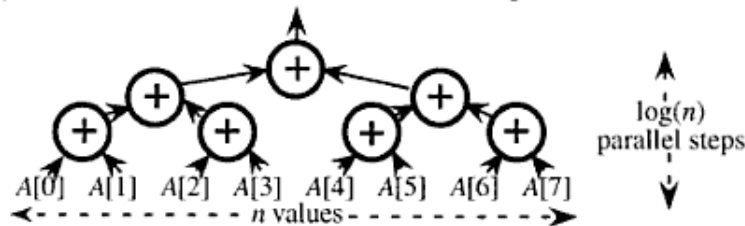
Exercise 12.4

- (a) Design a circuit family for the parity problem and describe it formally. Prove that there is a logspace uniform one.
- (b) Let $A[0..n]$ be an array of integers. Design a PRAM for summing numbers in an array, i.e. compute $\sum_{i=0}^n A[i]$.
Can you compute the array-suffix-sum, i.e. $\sum_{i=j}^n A[i]$ for all $0 \leq j \leq n$, with the same complexity?

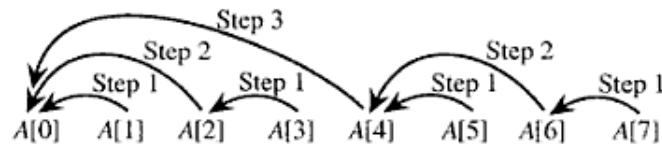
Solution:

Example: Summing an Array

Idea: To add up the items in an array $A[0] \dots A[n-1]$, in one step items one apart are added to cut the problem in half, in a second step items two apart are added to again cut the size of the problem in half, etc. This approach can be viewed as simulating a tree network with the array A stored in the leaves and the sum coming out of the root:



In practice, we do not really need all of the processors; we can keep overwriting A . Concurrent reads are not used, and the algorithm works in the EREW model.



The summing algorithm:

```

function erewSUM( $A[0] \dots A[n-1]$ )
  for  $k=0$  to  $\lceil \log_2(n) \rceil - 1$  do
    for  $0 \leq i < n - 2^{k+1}$  in parallel do
      if  $i$  is a multiple of  $2^{k+1}$  then  $A[i] := A[i] + A[i + 2^k]$ 
  return  $A[0]$ 
end
  
```

Note: The test if i is a multiple of 2^{k+1} is not necessary (see the exercises).

An equivalent way to express the summing algorithm:

```

function erewSUM( $A[0] \dots A[n-1]$ )
   $k := 1$ 
  while  $k < n$  do begin
    for  $0 \leq i < n - 2k$  in parallel do
      if  $i$  is a multiple of  $2k$  then  $A[i] := A[i] + A[i + k]$ 
     $k := k * 2$ 
  end
  return  $A[0]$ 
end
  
```

Complexity: Each of the $\lceil \log_2(n) \rceil$ iterations uses $O(1)$ time (since $O(1)$ time is used for the body of the parallel *for* loop); hence the algorithm is $O(\log(n))$ time. $O(1)$ space is used in addition to the space used by A . The number of processors used is $n/2$; however, we shall see later (Brent's Lemma) that $O(n/\log(n))$ processors suffice.

Example: List Prefix-Sum / List Ranking

Notation: L is a singly-linked list represented by $A[0] \dots A[n-1]$, $0 \leq first < n$ the index of the first item, and the array $NEXT[0] \dots NEXT[n-1]$ such that by starting with $i := A[first]$ and repeatedly doing $i := NEXT[i]$, we visit all positions and end up at a position i such that $NEXT[i] = nil$; for simplicity assume items are ≥ 0 and $nil = -1$.

List suffix-sum: We wish to compute for each $0 \leq i < n$ the sum of all positions from position i through the end of the list. We can use the same distance-doubling idea as for array sum, emanating from every vertex; only the EREW PRAM model is needed:

```

procedure erewListSuffixSum( $L$ )
  while  $NEXT[first] \neq nil$  do
    for  $0 \leq i < n$  in parallel do if  $NEXT[i] \neq nil$  then begin
       $A[i] := A[i] + A[NEXT[i]]$ 
       $NEXT[i] := NEXT[NEXT[i]]$ 
    end
  end

```

List prefix-sum: We wish to compute for each $0 \leq i < n$ the sum of all positions from position i through the start of the list. We can reverse the list (in parallel do $NEXT[NEXT[i]] := i$, set $first$ to what used to be the last position, and set the $NEXT$ field of what used to be the first position to nil) and then do a suffix sum (see the exercises).

List ranking: The special case of prefix-sum where all values of A are 1 (there could be additional data associated with each vertex), and we compute the position of each vertex.

Suffix-sum / prefix-sum / list ranking on an array: For the special case of a list in sequential positions of an array, define $NEXT[i] = i+1$, $0 \leq i < n-1$, and $NEXT[n-1] = nil$. For example, suppose the array $A[0] \dots A[9]$ initially contained 1 in each location and consider the successive iterations of the *while* loop of `erewListSuffixSum`:

array position	0	1	2	3	4	5	6	7	8	9
starting values	1	1	1	1	1	1	1	1	1	1
values after <i>first</i> iteration	2	2	2	2	2	2	2	2	2	1
values after <i>second</i> iteration	4	4	4	4	4	4	4	3	2	1
values after <i>third</i> iteration	8	8	8	7	6	5	4	3	2	1
values after <i>fourth</i> iteration	10	9	8	7	6	5	4	3	2	1

Complexity: $O(\log(n))$ time since each iteration of the outer *while* loop for suffix-sum doubles the distance over which sums are taken, and prefix-sum adds only $O(1)$ additional time. $O(n)$ space in addition to the space for L (or $O(n)$ additional space if we cannot overwrite A and $NEXT$ and must first make copies). $O(n)$ processors are used.

