CS 6810: Theory of Computing

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Lecturer: Eshan Chattopadhyay

Scribe: Omkar Bhalerao

## 1 NL = coNL

In the previous lecture, we introduced the following theorem:

**Theorem 1.1** (Immerman, Szelepscensyi). NL = coNL.

We continue by proving this theorem. To do so, we first define the language  $\overline{\text{PATH}}$ :

**Definition 1.2.**  $\overline{\text{PATH}} = \{ \langle G, s, t \rangle : \text{there is no s-t path in digraph } G = (V, E) \}.$ 

We now outline our proof statement along with some corollaries.

**Theorem 1.3.** There exists an  $O(\log n)$  space nondeterministic algorithm for  $\overline{\text{PATH}}$ .

Corollary 1.4. NL = coNL.

Corollary 1.5. For all "nice"  $S(n) > \log n$ , NSPACE(S(n)) = coNSPACE(S(n)).

As an aside, note that one implication of Corollary 1.5 is that NPSPACE = coNPSPACE; however, we have already shown this, since we know that NPSPACE = PSPACE and that PSPACE is closed under its complement. In addition, we contrast this theorem to Savitch's theorem from the previous lecture: that NSPACE(S(n))  $\subseteq$  DSPACE( $S(n)^2$ ). In other words, there's no overhead when going from NSPACE to coNSPACE, unlike when going from NSPACE to DSPACE.

To begin the proof, we have the following definition:

**Definition 1.6.**  $C_i = \{v \in V : there exists a path of length at most i from s to v\}.$ 

As such,  $\overline{\text{PATH}}$  is equivalent to asking whether  $t \notin C_n$ .

**Claim 1.7.** There exists an  $O(\log n)$  nondeterministic algorithm that can decide  $[v \in C_i]$  for all *i*.

The proof of this is similar to the proof given in the previous lecture for deciding PATH, but we will outline it below:

*Proof.* We will describe such an algorithm below:

On the work tape, have a section representing a counter, a section representing the current guess, and a section representing the next guess. Begin by writing s to the current guess section. Then, at each iteration, nondeterministically choose a neighbor u of s, write it to the next guess section, and increment the counter. Then, write the next guess to the current guess section, and repeat. If we see node v at any point in this process, accept; otherwise, after the counter reaches i, reject.

Since at most  $O(\log n)$  bits are needed to store the counter and the labels of all the nodes, this algorithm indeed only uses  $O(\log n)$  space.

**Claim 1.8.** Given  $|C_{i-1}| = r$ , there exists a nondeterministic  $O(\log n)$  space algorithm that can decide  $[v \notin C_i]$  for all *i*.

*Proof.* We describe such an algorithm below:

First, our NDTM will nondeterministically guess a set  $T \subseteq V$  such that |T| = r. Then, the machine will use claim 1 on all nodes in its guess to check whether it is correct. Though in general representing the set T could take space larger than  $O(\log n)$ , it's possible to sequentially go through each element of T to ensure that no more than  $O(\log n)$  space in total is used. Once a correct guess T is found, our NDTM will simply sequentially go through all the nodes in T and check if v is connected to each of the nodes. If v is connected to any one of the nodes, our algorithm will reject. Otherwise, it will accept.

**Claim 1.9.** Given  $|C_{i-1}| = r_{i-1}$ , there exists a nondeterministic  $O(\log n)$  space algorithm that either rejects, or outputs the correct size of  $C_i$ , and the said output happens for at least one execution path.

*Proof.* Again, we describe an algorithm.

For every node in the neighbors of  $C_{i-1}$ , our algorithm will nondeterministically guess whether  $v \in C_i$  or  $v \notin C_i$ , and then verifies those guesses using Claim 1 and Claim 2. It will also maintain a counter that starts at  $r_{i-1}$ . If the algorithm makes a correct guess, then either 1 or 0 is added to the counter. Since the Claim 1 and Claim 2 algorithms are guaranteed to be correct for at least one path of execution, it's therefore the case that there's at least one path of execution for which the Claim 3 algorithm is also correct. In addition, we will use the same sequential strategies as previously to ensure that no more than  $O(\log n)$  total space is used.

Now that we have these three claims, our overall algorithm to solve PATH is as follows: first, we set  $|c_0| = 1$ . From this, we compute  $|c_1|, \ldots, |c_n|$  sequentially using Claim 3 (implicitly using Claims 1 and 2 in the process). Lastly, we answer  $[t \notin C_n]$  using Claim 2. Thus, we've shown that  $\overline{\text{PATH}} \in \mathsf{NL}$  and thus that  $\mathsf{NL} = \mathsf{coNL}$ .

## 2 **PSPACE**-completeness

We now switch topics to PSPACE-completeness.

**Definition 2.1** (PSPACE-completeness). A language L is PSPACE-complete if  $L \in PSPACE$  and for all  $L' \in PSPACE$ ,  $L' \leq_P L$ , where  $\leq_P$  means that L' is polynomial-time Karp reducible to L.

To aid us in talking about PSPACE-completeness, we will now introduce a canonical language to represent PSPACE:

**Definition 2.2** (True Quantified Boolean Formulas (TQBF)). TQBF is the language of true quantified Boolean formulas; in other words, formulas of the following form that evaluate to true:

$$Q_1 x_1 Q_2 x_2 \dots Q_n x_n \phi(x_1, x_2, \dots, x_n)$$

where each  $Q_i$  is a quantifier (i.e.  $\forall$  or  $\exists$ ), and  $\phi$  is a Boolean formula of variables  $x_1, \ldots, x_n$ .

We will show that TQBF is indeed PSPACE-complete.

Claim 2.3. TQBF  $\in$  PSPACE.

*Proof.* We describe a polynomial-space algorithm for deciding TQBF.

First, our TM will brute-force over all variable assignments. It iterates over all quantifiers in the process, and in doing so, will ensure that for all  $\forall$ -quantified variables, both assignments result in a true formula, and for all  $\exists$ -quantified variables, at least one assignment results in a true formula. If this is true for all variables, our TM accepts; otherwise, it rejects.

This TM uses polynomial space because space can be reused across the different recursive calls. This is captured in the following recurrence, where n is the number of quantifiers, m is the size of the formula  $\phi$ , and S(n,m) is the space complexity at that level:

$$S(n,m) = S(n-1,m) + O(\operatorname{poly}(n,m))$$

Again, since we reuse space across each recursive call, this recurrence is correct. Thus, the total space complexity used is still O(poly(n, m)), proving that TQBF  $\in$  PSPACE.

## Claim 2.4. For all $L \in \mathsf{PSPACE}$ , $L \leq_P \mathsf{TQBF}$ .

*Proof.* For the sake of time, we omit the full proof; this will be given in the next lecture. We will, however, go through the setup steps for it.

First, note that if  $L \in \mathsf{PSPACE}$ , then there exists a TM M that uses  $cn^c$  space for some constant c and computes L. We will denote  $cn^c$  by S(n) from here on out.

We define  $G_{M,x}$  to be the configuration graph of M on input x. Then, by definition,  $x \in L$  iff there exists a path from  $c_{\text{start}}$  to  $c_{\text{accept}}$  in  $G_{M,x}$ , where  $c_{\text{start}}$  is the starting configuration and  $c_{\text{accept}}$ is the accepting configuration (WLOG assume that there's exactly one accepting configuration).

We now define a formula  $\phi_i(A, B) = 1$  for two configurations A, B if there exists a path from A to B in  $G_{M,x}$  of length at most  $2^i$ . Precisely, we take the Boolean variables that encode the configuration state of A and B and define a formula for which the above fact is true. Let  $\ell$  be the total number of variables in A and B; then, the formula is one on  $2\ell$  variables, where  $\ell \in O(S(n))$  by definition.

Clearly, we have that  $\phi_0(A, B) = 1$  iff  $(A, B) \in E(G_{M,x})$ , where E(G) represents the edge set of the graph G. As such,  $\phi_0$  can be encoded by the transition function of M. In the more general case where  $i \neq 0$ , we can simply write  $\phi_i(A, B) = (\exists C.\phi_{i-1}(A, C) \land \phi_{i-1}(C, B))$ . However, we're not quite done yet, because the formula  $\phi_i(A, B)$  could be potentially exponentially sized in A and B. There is, however, a trick to ensure that  $\phi_i(A, B)$  is polynomially sized, and we will cover that in the next lecture.