

## Savitch's Theorem

Last time (before the exam) we proved a few theorems in and around NP. We proved the Cook-Levin theorem that SAT was NP complete. We also proved Ladner's theorem, that if  $P \neq NP$  then there exist languages  $L \in NP$  and not NP-complete.

Today's lecture will be on space, that "other" resource. Space is a very different resource than time. After an algorithm finishes, you get the space back. You can never get the time back. This makes space a ~~more~~ less interesting resource to study. This also lets us use techniques and tricks which would not work for time.

Recall  $TIME(f(n))$ ,  $SPACE(f(n))$  are the classes of languages decidable in  $f(n)$  time/space. Note the following chain

$$TIME(f(n)) \subseteq SPACE(f(n)) \subseteq TIME(2^{O(f(n))})$$

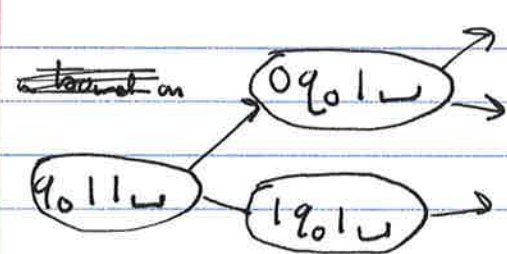
- Consider a language decidable in  $f(n)$  time. There exists a Turing machine which takes  $f(n)$  steps to decide this language. At each step, it may use at most one new cell of the tape. So a machine which uses  $f(n)$  time can use no more than  $f(n)$  space.

- We show a stronger result for the second one, with  $SPACE(P(n)) \subseteq NSPACE(P(n)) \subseteq TIME(2^{O(P(n))})$

The first containment following from the generalization of nondeterminism.

We now show the second containment in a creative way.

Given a language decidable by a nondeterministic Turing machine in  $f(n)$  space, we want to show this language is decidable deterministically in  $2^{O(f(n))}$  time. We will do so by graph search! For some specific  $N, w$ , let the configuration graph  $G$  be a directed graph such that each node corresponds to a configuration of  $N$  on  $w$ . Note that if  $N$  runs in  $f(n)$  space, then this graph has a bound on the possible number of ~~edges~~ vertices.



Note that we do not count the input as part of the space used. Sometimes the input is on a separate write

only tape. Since our machine  $N$  is non-deterministic, our graph may have arity greater than one. We may assume it has arity no more than ~~two~~ two. In order to show  $NSPACE(f(n)) \subseteq TIME(2^{O(f(n))})$  we give our  $2^{O(f(n))}$  time deterministic algorithm. First using  $N$ , we build the configuration graph. Then we BFS from the start configuration  $C_0$  to an accepting one  $C_a$ . BFS is linear time in the size of the input, this graph has worst case  $2^{O(f(n))}$  nodes. It also takes this long to build it, so we see this is a time  $2^{O(f(n))}$  deterministic algorithm. So  $NSPACE(f(n)) \subseteq TIME(2^{O(f(n))})$ .

Savitch's Theorem, our main result today:

$$NSPACE(f(n)) \subseteq SPACE(f(n)^2)$$

with some conditions on  $f$ . When hold on. Let's first interpret this result. We somehow are able to "determinize" something with only polynomial overhead. Could such a technique apply to P vs NP? Probably not, or someone would have found it by now. So although we only get a polynomial space cost, we probably get a super-polynomial, or exponential <sup>time</sup> ~~space~~ cost. Our deterministic algorithm may only use  $f^2(n)$  space, but it should use  $2^{f(n)}$  time.

A second immediate remark is that since polynomials are closed under composition, multiplication, we see  $NSPACE = PSPACE$ . This would, the study of space, already looks very different. This is analogous to P vs NP, but unlike that problem, the result is unexpected, and we ~~can~~ have been able to solve it. Now onto the proof.



Rather than some naive strategy, we are going to employ divide and conquer. We want to simulate a nondeterministic TM  $N$  which uses  $f(n)$  space, deterministically using no more than  $f^2(n)$  space. If  $C_0$  is the start,  $C_a$  is the accept configurations, and  $C$  is some other configuration, notice that  $C_0 \stackrel{*}{\vdash} C_a$  in  $t$  steps if  $C_0 \stackrel{*}{\vdash} C$  in  $t/2$  steps and  $C \stackrel{*}{\vdash} C_a$  in  $t/2$  steps for some  $t$ .

$$C_0 \overset{t}{\dashv} C_a \equiv C_0 \overset{t/2}{\dashv} C \overset{t/2}{\dashv} C_a$$

This will be our divide-and-conquer recurrence. Importantly, our recursive calls are run sequentially and reuse space.

def YIELDS( $C_i, C_j, t$ ):

if  $C_i = C_j$ : return true

if  $C_i \stackrel{(t=1)}{\vdash} C_j$ : in one step by the rules of  $\delta$  of  $N$  return true

if  $t > 1$ :

foreach  $C$  configuration of  $N$  using  $f(n)$  space:

YIELDS( $C_i, C, t/2$ )

YIELDS( $C, C_j, t/2$ )

if both accept, accept

reject

$M$  on input  $w$ :

$C_0$  = start configuration of  $N$

$C_a$  = accept configuration after tape head clears tape

$d$  = chosen so  $N$  has no more than  $2^{d \cdot f(n)}$  configurations

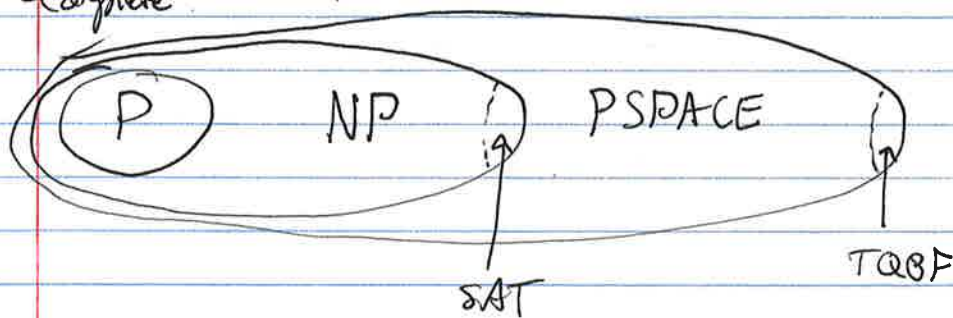
return YIELDS( $C_0, C_a, 2^{d \cdot f(n)}$ )

It certainly is correct.  $M$  is a deterministic simulator of  $N$ , so it decides the same language. Now onto the analysis.

We hope to show  $M$  simulates  $N$  using no more than  $f^2(n)$  space. For each recursive call, a stack frame containing  $C_i, C_j, t$  is stored. Each level of the recursion uses  $O(f(n))$  space, as worst case,  $C_i, C_j$  are  $O(f(n))$  since  $N$  uses  $f(n)$  space. Each level of the recursion divides  $t = 2^{df(n)}$  in half. If you recall anything about the Master theorem, or even geometric series, the depth of our recursion tree is  $\log t = \log 2^{df(n)} = O(f(n))$ . Since each level of the recursion takes  $O(f(n))$  space, and our recursion has  $O(f(n))$  depth, we observe the total space used is  $O(f(n)) \cdot O(f(n)) = O(f^2(n))$ .

I mentioned there were some restrictions on  $f(n)$ . First is that we may assume it is space-constructible, that  $M$  can compute  $f(n)$  within  $O(f(n))$  space. Most obvious functions have this property, but some crazy ones do not. Second is that  $f$  was super-linear. That  $f(n) \gg n$ . This can be improved to  $f(n) \geq \log n$  with some automata specification. A final remark, Hartmanis came up with a similar idea, but to prove a theorem about context-free languages.

Recall SAT is NP-complete, a boolean formula is like  $(x_1 \vee x_2 \vee x_3)$  right. This is not a boolean so much as it is a logical formula! we just hide the quantifiers, like  $\exists x_1 \exists x_2 \exists x_3 (x_1 \vee x_2 \vee x_3)$ . What if we allow for any quantifiers? Like  $\forall x_1 \forall x_2 \exists x_3 (x_1 \vee x_2 \vee x_3)$ ? This is called TQBF, True ~~quantified~~ quantified boolean formula.  $TQBF = \{ \phi \mid \phi \text{ is a true quantified boolean formula} \}$ . It was out how SAT is NP-complete, TQBF is PSPACE-complete.



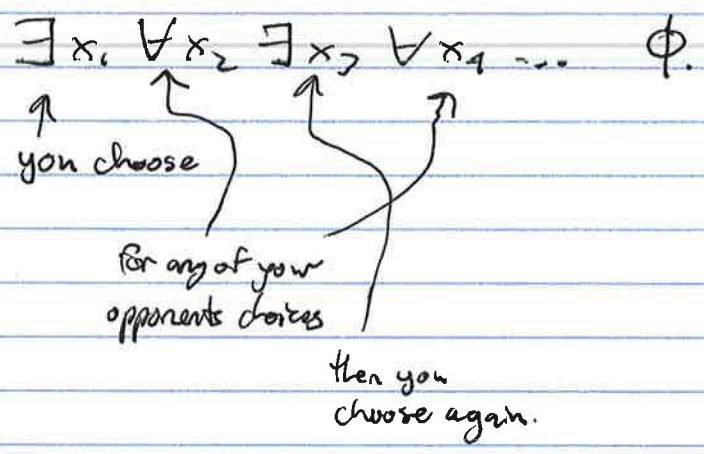


Notice SAT has structure like most puzzles. A puzzle, is a single-player device in which you make a sequence of decisions to reach some goal. Intuitively,  $\exists x, \exists x_2, \dots$  is you making your decisions. Most puzzles are then NP-complete.

Notice TQBF has structure like two player games of perfect information. Consider a TQBF like  $\exists \exists \forall \forall \exists \exists \dots$

you can compress this into like  $\exists \forall \exists \forall \dots$  right?  
 $\exists x_1, \exists x_2 \equiv \exists (x_1, x_2)$  with a little abuse of types.

Playing a TQBF with alternating quantifiers looks like a game!  
 Its quite literally minimax.



and generalizations

Most two player games, under appropriate restrictions, are PSPACE-complete. Chess, checkers, Go, and more. Appropriate restrictions would be that the game require perfect information (no shadowed areas of the map) and a polynomial bound on the depth of the number of moves. Without this bound many of these games are actually EXPTIME-complete, although these proofs are less general.

