

Complements of Nondeterministic Classes

- From p. 133, we know R, RE, and coRE are distinct.
 - coRE contains the complements of languages in RE, *not* the languages not in RE.
- Recall that the **complement** of L , denoted by \bar{L} , is the language $\Sigma^* - L$.
 - SAT COMPLEMENT is the set of unsatisfiable boolean expressions.
 - HAMILTONIAN PATH COMPLEMENT is the set of graphs without a Hamiltonian path.

The Co-Classes

- For any complexity class \mathcal{C} , $\text{co}\mathcal{C}$ denotes the class

$$\{L : \bar{L} \in \mathcal{C}\}.$$

- Clearly, if \mathcal{C} is a *deterministic* time or space *complexity class*, then $\mathcal{C} = \text{co}\mathcal{C}$.
 - They are said to be **closed under complement**.
 - A deterministic TM deciding L can be converted to one that decides \bar{L} within the same time or space bound by reversing the “yes” and “no” states.
- Whether nondeterministic classes for time are closed under complement is not known (p. 79).

Comments

- As

$$\text{co}\mathcal{C} = \{L : \bar{L} \in \mathcal{C}\},$$

$L \in \mathcal{C}$ if and only if $\bar{L} \in \text{co}\mathcal{C}$.

- But it is *not* true that $L \in \mathcal{C}$ if and only if $L \notin \text{co}\mathcal{C}$.
 - $\text{co}\mathcal{C}$ is not defined as $\bar{\mathcal{C}}$.
- For example, suppose $\mathcal{C} = \{\{2, 4, 6, 8, 10, \dots\}\}$.
- Then $\text{co}\mathcal{C} = \{\{1, 3, 5, 7, 9, \dots\}\}$.
- But $\bar{\mathcal{C}} = 2^{\{1,2,3,\dots\}^*} - \{\{2, 4, 6, 8, 10, \dots\}\}$.

The Quantified Halting Problem

- Let $f(n) \geq n$ be proper.
- Define

$$H_f = \{M; x : M \text{ accepts input } x \\ \text{after at most } f(|x|) \text{ steps}\},$$

where M is deterministic.

- Assume the input is binary.

$$H_f \in \text{TIME}(f(n)^3)$$

- For each input $M; x$, we simulate M on x with an alarm clock of length $f(|x|)$.
 - Use the single-string simulator (p. 57), the universal TM (p. 118), and the linear speedup theorem (p. 64).
 - Our simulator accepts $M; x$ if and only if M accepts x before the alarm clock runs out.
- From p. 63, the total running time is $O(\ell_M k_M^2 f(n)^2)$, where ℓ_M is the length to encode each symbol or state of M and k_M is M 's number of strings.
- As $\ell_M k_M^2 = O(n)$, the running time is $O(f(n)^3)$, where the constant is independent of M .

$$H_f \notin \text{TIME}(f(\lfloor n/2 \rfloor))$$

- Suppose TM M_{H_f} decides H_f in time $f(\lfloor n/2 \rfloor)$.
- Consider machine $D_f(M)$:

if $M_{H_f}(M; M) = \text{“yes”}$ **then** “no” **else** “yes”

- D_f on input M runs in the same time as M_{H_f} on input $M; M$, i.e., in time $f(\lfloor \frac{2n+1}{2} \rfloor) = f(n)$, where $n = |M|$.^a

^aA student pointed out on October 6, 2004, that this estimation omits the time to write down $M; M$.

The Proof (concluded)

- First,

$$D_f(D_f) = \text{“yes”}$$

$$\Rightarrow D_f; D_f \notin H_f$$

$$\Rightarrow D_f \text{ does not accept } D_f \text{ within time } f(|D_f|)$$

$$\Rightarrow D_f(D_f) \neq \text{“yes”}$$

$$\Rightarrow D_f(D_f) = \text{“no”}$$

a contradiction

- Similarly, $D_f(D_f) = \text{“no”} \Rightarrow D_f(D_f) = \text{“yes.”}$

The Time Hierarchy Theorem

Theorem 16 *If $f(n) \geq n$ is proper, then*

$$\text{TIME}(f(n)) \subsetneq \text{TIME}(f(2n + 1)^3).$$

- The quantified halting problem makes it so.

Corollary 17 $P \subsetneq \text{EXP}$.

- $P \subseteq \text{TIME}(2^n)$ because $\text{poly}(n) \leq 2^n$ for n large enough.
- But by Theorem 16,

$$\text{TIME}(2^n) \subsetneq \text{TIME}((2^{2n+1})^3) \subseteq \text{TIME}(2^{n^2}) \subseteq \text{EXP}.$$

- So $P \subsetneq \text{EXP}$.

The Space Hierarchy Theorem

Theorem 18 (Hennie and Stearns (1966)) *If $f(n)$ is proper, then*

$$\text{SPACE}(f(n)) \subsetneq \text{SPACE}(f(n) \log f(n)).$$

Corollary 19 $L \subsetneq \text{PSPACE}$.

Nondeterministic Time Hierarchy Theorems

Theorem 20 (Cook (1973)) *If $f(n)$ is proper, then*

$$\text{NTIME}(n^r) \subsetneq \text{NTIME}(n^s)$$

whenever $1 \leq r < s$.

Theorem 21 (Seiferas, Fischer, and Meyer (1978)) *If $T_1(n), T_2(n)$ are proper, then*

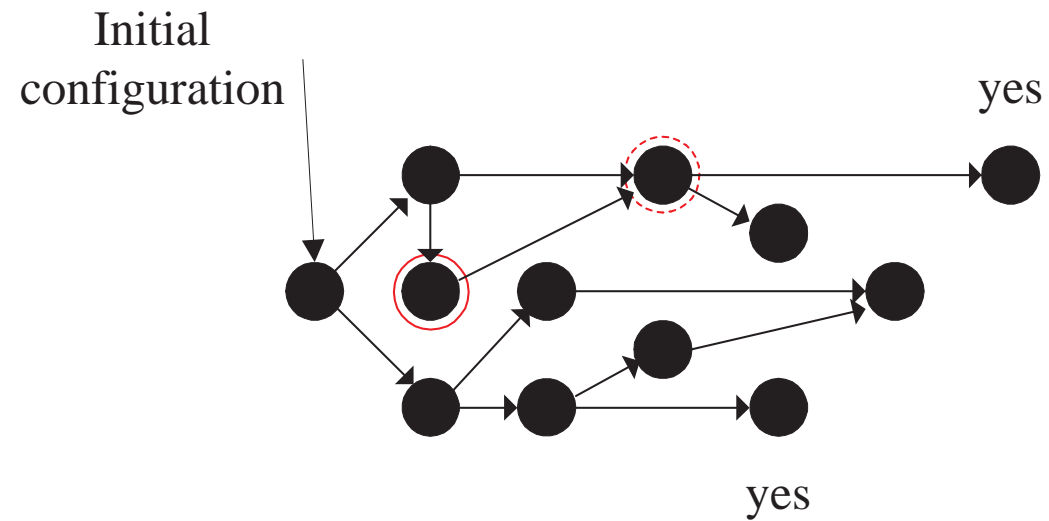
$$\text{NTIME}(T_1(n)) \subsetneq \text{NTIME}(T_2(n))$$

whenever $T_1(n+1) = o(T_2(n))$.

The Reachability Method

- The computation of a time-bounded TM can be represented by a directed graph.
- The TM's configurations constitute the nodes.
- Two nodes are connected by a directed edge if one yields the other.
- The start node representing the initial configuration has zero in degree.
- When the TM is nondeterministic, a node may have an out degree greater than one.

Illustration of the Reachability Method



Relations between Complexity Classes

Theorem 22 *Suppose $f(n)$ is proper. Then*

1. $\text{SPACE}(f(n)) \subseteq \text{NSPACE}(f(n))$,
 $\text{TIME}(f(n)) \subseteq \text{NTIME}(f(n))$.
 2. $\text{NTIME}(f(n)) \subseteq \text{SPACE}(f(n))$.
 3. $\text{NSPACE}(f(n)) \subseteq \text{TIME}(k^{\log n + f(n)})$.
- Proof of 2:
 - Explore the computation *tree* of the NTM for “yes.”
 - Specifically, generate a $f(n)$ -bit sequence denoting the nondeterministic choices over $f(n)$ steps.

Proof of Theorem 22(2)

- (continued)
 - Simulate the NTM based on the choices.
 - Recycle the space and then repeat the above steps until a “yes” is encountered or the tree is exhausted.
 - Each path simulation consumes at most $O(f(n))$ space because it takes $O(f(n))$ time.
 - The total space is $O(f(n))$ because space is recycled.

Proof of Theorem 22(3)

- Let k -string NTM

$$M = (K, \Sigma, \Delta, s)$$

with input and output decide $L \in \text{NSPACE}(f(n))$.

- Use the reachability method on the configuration graph of M on input x of length n .
- A configuration is a $(2k + 1)$ -tuple

$$(q, w_1, u_1, w_2, u_2, \dots, w_k, u_k).$$

Proof of Theorem 22(3) (continued)

- We only care about

$$(q, i, w_2, u_2, \dots, w_{k-1}, u_{k-1}),$$

where i is an integer between 0 and n for the position of the first cursor.

- The number of configurations is therefore at most

$$|K| \times (n + 1) \times |\Sigma|^{(2k-4)f(n)} = O(c_1^{\log n + f(n)}) \quad (1)$$

for some c_1 , which depends on M .

- Add edges to the configuration graph based on M 's transition function.

Proof of Theorem 22(3) (concluded)

- $x \in L \Leftrightarrow$ there is a path in the configuration graph from the initial configuration to a configuration of the form (“yes”, i, \dots) [there may be many of them].
- This is REACHABILITY on a graph with $O(c_1^{\log n + f(n)})$ nodes.
- It is in $\text{TIME}(c^{\log n + f(n)})$ for some c because $\text{REACHABILITY} \in \text{TIME}(n^j)$ for some j and

$$\left[c_1^{\log n + f(n)} \right]^j = (c_1^j)^{\log n + f(n)}.$$

Space-Bounded Computation and Proper Functions

- In the definition of *space-bounded* computations earlier, the TMs are not required to halt at all.
- When the space is bounded by a proper function f , computations can be assumed to halt:
 - Run the TM associated with f to produce an output of length $f(n)$ first.
 - The space-bounded computation must repeat a configuration if it runs for more than $c^{\log n + f(n)}$ steps for some c (p. 196).
 - So we can prevent infinite loops during simulation by pruning any path longer than $c^{\log n + f(n)}$.

The Grand Chain of Inclusions

$$L \subseteq NL \subseteq P \subseteq NP \subseteq PSPACE \subseteq EXP.$$

- By Corollary 19 (p. 189), we know $L \subsetneq PSPACE$.
- The chain must break somewhere between L and $PSPACE$.^a
- It is suspected that all four inclusions are proper.
- But there are no proofs yet.^b

^aBill Gates (1996), “I keep bumping into that silly quotation attributed to me that says 640K of memory is enough.”

^bCarl Friedrich Gauss (1777–1855), “I could easily lay down a multitude of such propositions, which one could neither prove nor dispose of.”

Nondeterministic Space and Deterministic Space

- By Theorem 4 (p. 84),

$$\text{NTIME}(f(n)) \subseteq \text{TIME}(c^{f(n)}),$$

an exponential gap.

- There is no proof yet that the exponential gap is inherent.
- How about NSPACE vs. SPACE?
- Surprisingly, the relation is only quadratic—a polynomial—by Savitch's theorem.

Savitch's Theorem

Theorem 23 (Savitch (1970))

REACHABILITY \in SPACE($\log^2 n$).

- Let $G(V, E)$ be a graph with n nodes.
- For $i \geq 0$, let

PATH(x, y, i)

mean there is a path from node x to node y of length at most 2^i .

- There is a path from x to y if and only if

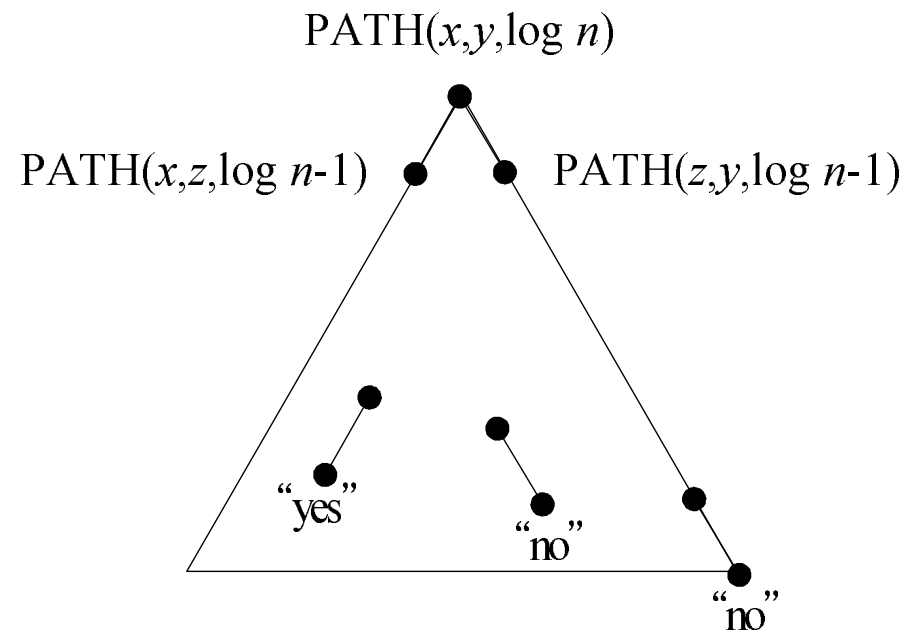
PATH($x, y, \lceil \log n \rceil$)

holds.

The Proof (continued)

- For $i > 0$, $\text{PATH}(x, y, i)$ if and only if there exists a z such that $\text{PATH}(x, z, i - 1)$ and $\text{PATH}(z, y, i - 1)$.
- For $\text{PATH}(x, y, 0)$, check the input graph or if $x = y$.
- Compute $\text{PATH}(x, y, \lceil \log n \rceil)$ with a depth-first search on a graph with nodes (x, y, z, i) s (see next page).^a
- Like stacks in recursive calls, we keep only the current path of (x, y, i) s.
- The space requirement is proportional to the depth of the tree: $\lceil \log n \rceil$.

^aContributed by Mr. Chuan-Yao Tan on October 11, 2011.



- Depth is $\lceil \log n \rceil$, and each node (x, y, z, i) needs space $O(\log n)$.
- The total space is $O(\log^2 n)$.

The Proof (concluded): Algorithm for $\text{PATH}(x, y, i)$

```
1: if  $i = 0$  then  
2:   if  $x = y$  or  $(x, y) \in E$  then  
3:     return true;  
4:   else  
5:     return false;  
6:   end if  
7: else  
8:   for  $z = 1, 2, \dots, n$  do  
9:     if  $\text{PATH}(x, z, i - 1)$  and  $\text{PATH}(z, y, i - 1)$  then  
10:      return true;  
11:    end if  
12:  end for  
13:  return false;  
14: end if
```


The Relation between Nondeterministic Space and Deterministic Space Only Quadratic

Corollary 24 *Let $f(n) \geq \log n$ be proper. Then*

$$\text{NSPACE}(f(n)) \subseteq \text{SPACE}(f^2(n)).$$

- Apply Savitch's proof to the configuration graph of the NTM on the input.
- From p. 196, the configuration graph has $O(c^{f(n)})$ nodes; hence each node takes space $O(f(n))$.
- But if we construct explicitly the whole graph before applying Savitch's theorem, we get $O(c^{f(n)})$ space!

The Proof (continued)

- The way out is *not* to generate the graph at all.
- Instead, keep the graph implicit.
- We check for connectedness only when $i = 0$ on p. 204, by examining the input string G .
- There, given configurations x and y , we go over the Turing machine's program to determine if there is an instruction that can turn x into y in one step.^a

^aThanks to a lively class discussion on October 15, 2003.

The Proof (concluded)

- The z variable in the algorithm on p. 204 simply runs through all possible valid configurations.
 - Let $z = 0, 1, \dots, O(c^{f(n)})$.
 - Make sure z is a valid configuration before using it in the recursive calls.^a
- Each z has length $O(f(n))$ by Eq. (1) on p. 196.

^aThanks to a lively class discussion on October 13, 2004.

Implications of Savitch's Theorem

- $PSPACE = NPSPACE$.
- Nondeterminism is less powerful with respect to space.
- Nondeterminism may be very powerful with respect to time as it is not known if $P = NP$.

Nondeterministic Space Is Closed under Complement

- Closure under complement is trivially true for deterministic complexity classes (p. 182).
- It is known that^a

$$\text{coNSPACE}(f(n)) = \text{NSPACE}(f(n)). \quad (2)$$

- So

$$\text{coNL} = \text{NL},$$

$$\text{coNPSPACE} = \text{NPSPACE}.$$

- But there are still no hints of $\text{coNP} = \text{NP}$.

^aSzelepcényi (1987) and Immerman (1988).

Reductions and Completeness

Degrees of Difficulty

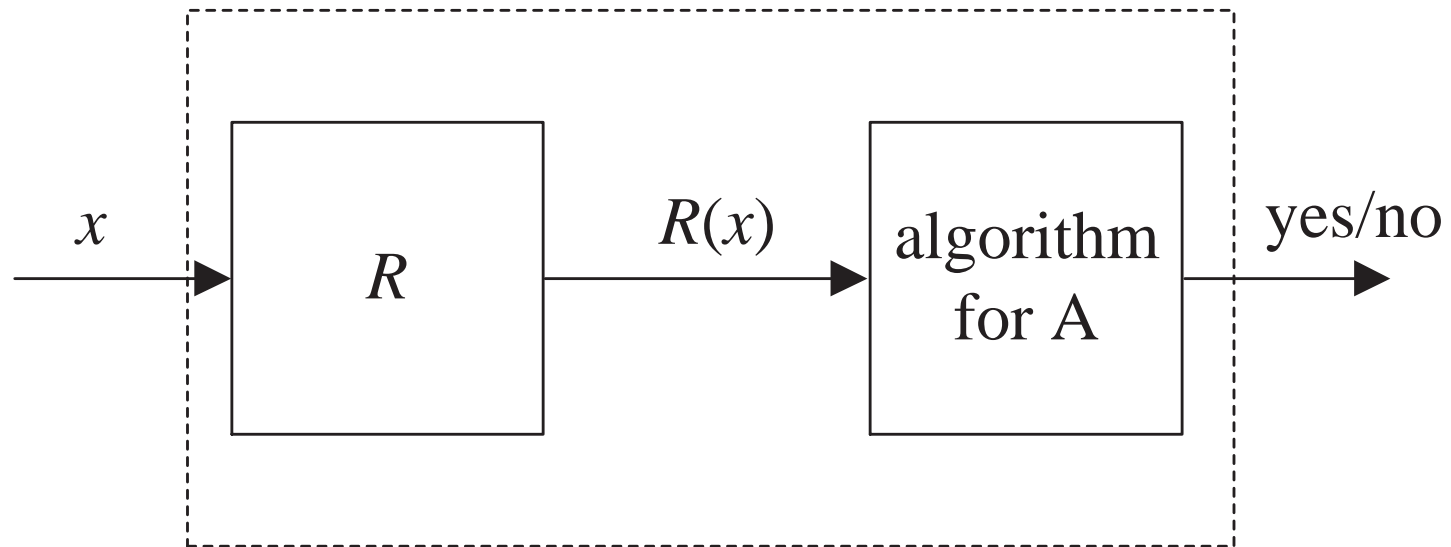
- When is a problem more difficult than another?
- **B reduces to A** if there is a transformation R which for every input x of B yields an equivalent input $R(x)$ of A.
 - The answer to x for B is the same as the answer to $R(x)$ for A.
 - There must be restrictions on the complexity of computing R .
 - Otherwise, $R(x)$ may solve B, defeating the purpose.
 - * E.g., $R(x) = \text{“yes”}$ if and only if $x \in B$!

Degrees of Difficulty (concluded)

- We say problem A is at least as hard as problem B if B reduces to A .
- This makes intuitive sense: If A is able to solve your problem B after only a little bit of work of R , then A must be at least as hard.
 - If A is easy to solve, it combined with R (which is also easy) would make B easy to solve, too.^a
 - If B is hard to solve, A must be hard (if not harder) to solve, too.

^aThanks to a lively class discussion on October 13, 2009.

Reduction



Solving problem B by calling the algorithm for problem A *once* and *without* further processing its answer.

Comments^a

- Suppose B reduces to A via a transformation R .
- The input x is an instance of B .
- The output $R(x)$ is an instance of A .
- $R(x)$ may not span all possible instances of A .^b
 - Some instances of A may never appear in the range of R .

^aContributed by Mr. Ming-Feng Tsai (D92922003) on October 29, 2003.

^b $R(x)$ may not be onto; Mr. Alexandr Simak (D98922040) on October 13, 2009.

Reduction between Languages

- Language L_1 is **reducible to** L_2 if there is a function R computable by a deterministic TM in space $O(\log n)$.
- Furthermore, for all inputs x , $x \in L_1$ if and only if $R(x) \in L_2$.
- R is said to be a **(Karp) reduction** from L_1 to L_2 .

Reduction between Languages (concluded)

- Note that by Theorem 22 (p. 193), R runs in polynomial time.
 - In most cases, a polynomial-time R suffices for proofs.
- Suppose R is a reduction from L_1 to L_2 .
- Then solving “ $R(x) \in L_2?$ ” is an algorithm for solving “ $x \in L_1?$ ”^a

^aBut it may not be an optimal one.

A Paradox?

- Degree of difficulty is not defined in terms of *absolute* complexity.
- So a language $B \in \text{TIME}(n^{99})$ may be “easier” than a language $A \in \text{TIME}(n^3)$.
 - This happens when B is reducible to A.
- But isn't this a contradiction if the best algorithm for B requires n^{99} steps?
- That is, how can a problem *requiring* n^{99} steps be reducible to a problem solvable in n^3 steps?

Paradox Resolved

- The so-called contradiction does not hold.
- When we solve the problem “ $x \in B?$ ” via “ $R(x) \in A?$ ”, we must consider the time spent by $R(x)$ and its length $|R(x)|$.
- If $|R(x)| = \Omega(n^{33})$, then answering “ $R(x) \in A?$ ” takes $\Omega((n^{33})^3) = \Omega(n^{99})$ steps, and there is no contradiction.
- Suppose, on the other hand, that $|R(x)| = o(n^{33})$.
- Then $R(x)$ must run in time $\Omega(n^{99})$ to make the overall time for answering “ $R(x) \in A?$ ” take $\Omega(n^{99})$ steps.
- In either case, the contradiction disappears.