Computational Complexity

Lecture 5: Space complexity

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Recap *What we saw last time..*

Limits of diagonalization, relativizing results

Oracles

• There exist $A, B \subseteq \{0, 1\}^*$ such that $P^A = NP^A$ and $P^B \neq NP^B$.

- Space-bounded computation
- Limits on memory space
- L, NL, PSPACE, NPSPACE
- Logspace reductions
- NL-completeness

- Instead of measuring the number T(n) of steps, we will measure the number S(n) of tape cells used
- For time bounds, T(n) < n typically makes no sense
 - In less than n steps, the machine cannot even read the input
- However, for space bounds, S(n) < n does make sense in some situations
- For space-bounded computation:
 - The input tape is read-only
 - We count how many tape cells on the 'work tapes' are used

Definition (SPACE)

Let $S : \mathbb{N} \to \mathbb{N}$ be a function. A decision problem $L \subseteq \Sigma^*$ is in SPACE(S(n)) if there exists a Turing machine that decides L and that on inputs of length n its tape heads (excluding on the input tape) visit at most $c \cdot S(n)$ tape cells.

Definition (NSPACE)

Let $S : \mathbb{N} \to \mathbb{N}$ be a function. A decision problem $L \subseteq \Sigma^*$ is in NSPACE(S(n)) if there exists a *nondeterministic* Turing machine that decides L and that on inputs of length n its tape heads (excluding on the input tape) visit at most $c \cdot S(n)$ tape cells.

Theorem If $S : \mathbb{N} \to \mathbb{N}$ is a space-constructible function, then: $DTIME(S(n)) \subseteq SPACE(S(n)) \subseteq NSPACE(S(n)) \subseteq DTIME(2^{O(S(n))}).$

- Assumption of space-constructibility rules out 'weird' functions.
 - S is space-constructible if there exists a TM that computes the function $x \mapsto S(|x|)$ in space O(S(|x|)), for each $x \in \{0, 1\}^*$

Some space classes

Definition

$$PSPACE = \bigcup_{c \ge 1} SPACE(n^c) \qquad L = SPACE(\log n)$$
$$NPSPACE = \bigcup_{c \ge 1} NSPACE(n^c) \qquad NL = NSPACE(\log n)$$

- By the previous theorem, then $L \subseteq NL \subseteq P$ and PSPACE \subseteq NPSPACE \subseteq EXP.
- What is an example of a problem in PSPACE? SAT
- What is an example of a problem in NL? Reachability in graphs

Theorem

If $f, g : \mathbb{N} \to \mathbb{N}$ are space-constructible functions such that f(n) is o(g(n)), then:

 $SPACE(f(n)) \subsetneq SPACE(g(n))$ and $NSPACE(f(n)) \subsetneq NSPACE(g(n))$.

• As a result: $L \subsetneq PSPACE$ and $NL \subsetneq NPSPACE$.

Definition (QBFs)

A quantified Boolean formula (QBF) (in prenex form) is of the form $Q_1x_1Q_2x_2\cdots Q_mx_m \varphi(x_1,\ldots,x_m)$, where each Q_i is one of the two quantifiers \exists or \forall , where the variables x_1,\ldots,x_m range over $\{0,1\}$, and where φ is a propositional formula (without quantifiers).

Truth of QBFs is defined recursively, based on the typical semantics of \exists and \forall .

• For example, $\exists x_1 \forall x_2 \ (x_1 \lor \neg x_2) \land (x_1 \lor x_2)$ is a QBF

Definition (TQBF)

The language TQBF consists of all QBFs that are true.

Theorem

TQBF is PSPACE-complete (under polynomial-time reductions).

- Why is TQBF in PSPACE?
 - Use a recursive algorithm.

For $\varphi = \exists x_i \ \psi$, recurse on $\psi[x_i \mapsto 0]$ and $\psi[x_i \mapsto 1]$, and return 1 if and only if at least one of the recursive calls returns 1. Similarly for $\varphi = \forall x_i \ \psi$.

- This takes exponential time, but polynomial space:
 - The recursion depth is linear in $|\varphi|$.
 - Space can be reused.
 - With polynomial space, we keep track of the position in the recursion tree, and if we're going up or down.





Theorem

TQBF is PSPACE-complete (under polynomial-time reductions).

- Why is TQBF PSPACE-hard?
 - Reduce arbitrary polynomial-space computation of TM M on input x to TQBF; (computation that uses p(n) space takes time at most 2^{q(n)})
 - Main idea: construct a QBF $\varphi_{c_1,c_2,t}$ that expresses that the computation leads from configuration c_1 to configuration c_2 within t steps, and return $\varphi_{c_0,c_{accent},2^{q(n)}}$
 - $\varphi_{c_1,c_2,t}$ has propositional variables that correspond to the configurations c_1,c_2
 - For t = 1, this can be done analogously to the proof of the Cook-Levin Theorem
 - For t > 1: $\varphi_{c_1,c_2,t}$ expresses $\exists m (\varphi_{c_1,m,\lceil t/2 \rceil} \land \varphi_{m,c_2,\lceil t/2 \rceil}) m$ is a sequence of vars
 - To avoid exponential blowup, write $\varphi_{c_1,c_2,t}$ in the following way:

$$\exists m \forall c_3 \forall c_4 \ ((``c_3 = c_1'' \land ``c_4 = m'') \lor (``c_3 = m'' \land ``c_4 = c_2'')) \rightarrow \varphi_{c_3, c_4, \lceil t/2 \rceil}$$



For every space-constructible $S : \mathbb{N} \to \mathbb{N}$ with $S(n) \ge \log n$:

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NSPACE(S(n)) \subseteq SPACE(S(n)^2).
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- So, in particular, PSPACE = NPSPACE.
- Proof strategy (for PSPACE = NPSPACE):
 - Show that TQBF is NPSPACE-complete and in PSPACE.

- To investigate $L \stackrel{?}{=} NL$, we need reductions that are weak enough.
- Since $L \subseteq NL \subseteq P$, every problem in $L \cup NL$ is reducible to each other using polynomial-time reductions.
 - You can solve any problem in $L \cup NL$ in polynomial time.
 - Reduction: solve the problem, and output a trivial yes-input or a trivial no-input.

Definition

A function $f : \{0,1\}^* \to \{0,1\}^*$ is implicitly logspace computable if:

- f is polynomially bounded, i.e., there exists some c such that $|f(x)| \le |x|^c$ for every $x \in \{0, 1\}^*$, and
- the languages $L_f = \{ (x, i) \mid f(x)_i = 1 \}$ and $L'_f = \{ (x, i) \mid i \leq |f(x)| \}$ are in the complexity class L, where $f(x)_i$ denotes the *i*th bit of f(x).

Definition

A language B is logspace-reducible to a language C (also written $B \leq_{\ell} C$) if there is a function $f : \{0,1\}^* \to \{0,1\}^*$ that is implicitly logspace computable and for each $x \in \{0,1\}^*$ it holds that $x \in B$ if and only if $f(x) \in C$.

- A language B is NL-complete if $B \in NL$ and $C \leq_{\ell} B$ for every $C \in NL$.
- Logspace reductions are transitive: if $B \leq_{\ell} C$ and $C \leq_{\ell} D$, then $B \leq_{\ell} D$.
- If $B \leq_{\ell} C$ and $C \in L$, then $B \in L$.

• So, if any NL-complete language is in L, then L = NL.

An NL-complete problem

Consider graph reachability in directed graphs:

 $\mathsf{PATH} = \{ (G, s, t) \mid G = (V, E) \text{ is a directed graph, } s, t \in V, \\ and t \text{ is reachable from } s \text{ in } G \}$

- PATH is NL-complete. Why is it in NL?
 - Keep the current and next node in memory (logspace).
 - Guess the next node, check if they are connected, and forget the previous node.
 - Start at *s*, accept if you reach *t*.
 - Keep the length of the path you already visited in memory (logspace), and stop when it is longer than |V| (to avoid looping forever).



Theorem (Immerman 1988, Szelepcsényi 1987)

For every space-constructible $S : \mathbb{N} \to \mathbb{N}$ with $S(n) > \log n$:

NSPACE(S(n)) = coNSPACE(S(n)).

• In particular: NL = coNL.

An overview of complexity classes



- Space-bounded computation
- Limits on memory space
- L, NL, PSPACE = NPSPACE
- Logspace reductions
- NL-completeness

- Complexity classes between P and PSPACE
- The Polynomial Hierarchy
- Bounded quantifier alternation
- Alternating Turing machines