# **Computational Complexity**

Lecture 6: 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  $P^B \neq NP^B$ .

# What will we do today?

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

### Space-bounded computation

- 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

### SPACE and NSPACE

### 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.

## Some first relations between time and space

#### Theorem

If  $S : \mathbb{N} \to \mathbb{N}$  is a space-constructible function, then:

$$\mathsf{DTIME}(S(n)) \subseteq \mathsf{SPACE}(S(n)) \subseteq \mathsf{NSPACE}(S(n)) \subseteq \mathsf{DTIME}(2^{O(S(n))}).$$

- 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

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

- 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

# Space Hierarchy Theorem

### Theorem

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

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

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

### Quantified Boolean Formulas

### 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  $\varphi$  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)

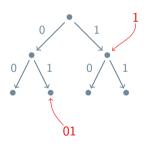
The language TQBF consists of all QBFs that are true.

### PSPACE-completeness

#### 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.



$$\{x_1\mapsto 0, x_2\mapsto 1\}$$

### Savitch's Theorem

### Theorem (Savitch 1970)

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

$$NSPACE(S(n)) \subseteq SPACE(S(n)^2).$$

■ So, in particular, PSPACE = NPSPACE.

- Proof strategy (for PSPACE = NPSPACE):
  - Show that TQBF is NPSPACE-complete and in PSPACE.

### Logspace reductions

- 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.
  - lacksquare 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.

# Logspace reductions (ct'd)

#### 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  $f(x) \in C$ .

### **NL-completeness**

- 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  $B <_{\ell} C$  and  $C \in L$ , then  $B \in L$ .

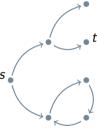
 $\blacksquare$  So, if any NL-complete language is in L, then L = NL.

### An NL-complete problem

■ Consider graph reachability in directed graphs:

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

- 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).



### Immerman-Szelepcsényi Theorem

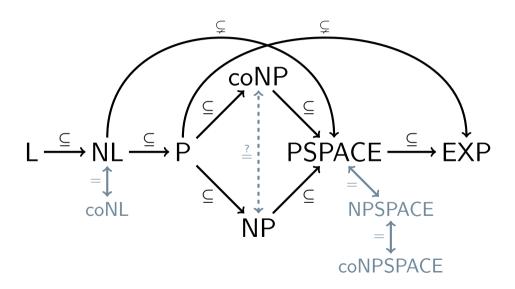
Theorem (Immerman 1988, Szelepcsényi 1987)

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



## Recap

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

### Next time

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