



### COMPLEXITY THEORY

**Lecture 11: Games/Logarithmic Space** 

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# Review

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### Review: PSpace-complete problems

We have encountered some PSpace-complete problems so far:

- The word problem for polynomially space bounded (N)TMs
- TRUE QBF
- FOL Model Checking (and SQL query answering)

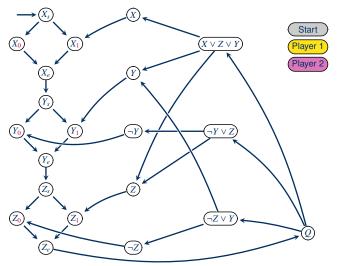
Several typical PSpace problems are related to the existence of winning strategies in 2-player games:

- FORMULA GAME
- GEOGRAPHY

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# Review: Geography is PSpace-hard

We consider the formula  $\exists X. \forall Y. \exists Z. (X \lor Z \lor Y) \land (\neg Y \lor Z) \land (\neg Z \lor Y)$ 



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#### More Games

The characteristic of PSpace is quantifier alternation

This is closely related to taking turns in 2-player games.

#### Are many games PSpace-complete?

- Issue 1: many games are finite that is: computationally trivial
  - → generalise games to arbitrarily large boards
    - generalised Tic-Tac-Toe is PSpace-complete
    - generalised Reversi is PSpace-complete
    - it is not always clear how to generalise a game (Generalised Backgammon?)
- Issue 2: (generalised) games where moves can be reversed may require very long matches
  - → such games often are even harder
    - generalised Go with Japanese ko rule is ExpTime-complete
    - generalised Draughts (Checkers) is ExpTime-complete
    - generalised Chess (without 50-move no-capture draw rule) is ExpTime-complete

Surprisingly, some of these games, e.g. Chess, are known to become even harder – namely ExpSpace-complete – if the exact same board position is not allowed to re-occur in a match. For Go, this case is open (link).

# Logarithmic Space

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### Logarithmic Space

### Polynomial space

As we have seen, polynomial space is already quite powerful.

We therefore consider more restricted space complexity classes.

### Linear space

Even linear space is enough to solve SAT.

### Sub-linear space

To get sub-linear space complexity, we consider Turing-machines with separate input tape and only count working space.

#### Recall:

$$\mathsf{L} = \mathsf{LogSpace} = \mathsf{DSpace}(\log n)$$

NL = NLogSpace = NSpace(log n)

### Problems in L and NL

What sort of problems are in L and NL?

In logarithmic space we can store

- a fixed number of counters (up to length of input)
- a fixed number of pointers to positions in the input string

#### Hence,

- L contains all problems requiring only a constant number of counters/pointers for solving.
- NL contains all problems requiring only a constant number of counters/pointers for verifying solutions.

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# Examples: Problems in L

**Example 11.1:** The language  $\{0^n1^n \mid n \ge 0\}$  is in L.

#### Algorithm:

- Check that no 1 is ever followed by a 0
   Requires no working space (only movements of the read head)
- Count the number of 0's and 1's
- Compare the two counters

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### Examples: Problems in L

#### **PALINDROMES**

Input: Word w on some input alphabet  $\Sigma$ 

Problem: Does w read the same forward and back-

ward?

#### Example 11.2: Palindromes $\in L$ .

#### Algorithm:

- Use two pointers, one to the beginning and one to the end of the input.
- At each step, compare the two symbols pointed to.
- Move the pointers one step inwards.

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### Example: A Problem in NL

#### REACHABILITY a.k.a. STCON a.k.a. PATH

Input: Directed graph G, vertices  $s, t \in V(G)$ 

Problem: Does G contain a path from s to t?

#### **Example 11.3: Reachability** ∈ NL.

#### Algorithm:

- Use a pointer to the current vertex, starting in s
- Iteratively move pointer from current vertex to some neighbour vertex nondeterministically
- Accept when finding t; reject when searching for too long

### An Algorithm for REACHABILITY

#### More formally:

```
01 CANREACH(G,s,t):
  c := |V(G)| // counter
03 p := s // pointer
04 while c > 0:
05 if p = t:
06 return TRUE
07 else:
08
       nondeterministically select G-successor p' of p
09
      p := p'
10 c := c - 1
11 // eventually, if no success:
12 return FALSE
```

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# Defining Reductions in Logarithmic Space

To compare the difficulty of problems in P or NL, polynomial-time reductions are useless. Recall the respective result from Lecture 5:

**Theorem 5.22:** If **B** is any language in P, **B**  $\neq \emptyset$ , and **B**  $\neq \Sigma^*$ , then **A**  $\leq_p$  **B** for any **A**  $\in$  P.

This also applies to languages in NL ( $\subseteq$  P).

**Definition 11.4:** A log-space transducer  $\mathcal{M}$  is a logarithmic space bounded Turing machine with a read-only input tape and a write-only, write-once output tape, and that halts on all inputs.

A log-space transducer  $\mathcal{M}$  computes a function  $f: \Sigma^* \to \Sigma^*$ , where f(w) is the content of the output tape of  $\mathcal{M}$  running on input w when  $\mathcal{M}$  halts.

In this case, *f* is called a log-space computable function.

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### Log-Space Reductions and NL-Completeness

**Definition 11.5:** A log-space reduction from  $\mathbf{L} \subseteq \Sigma^*$  to  $\mathbf{L}' \subseteq \Sigma^*$  is a log-space computable function  $f: \Sigma^* \to \Sigma^*$  such that for all  $w \in \Sigma^*$ :

$$w \in \mathbf{L} \iff f(w) \in \mathbf{L}'$$

We write  $\mathbf{L} \leq_L \mathbf{L}'$  in this case.

**Definition 11.6:** A problem  $L \in NL$  is complete for NL if every other language in NL is log-space reducible to L.

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Detour: P-completeness

Log-space reductions are also used to define P-complete problems:

**Definition 11.7:** A problem  $L \in P$  is complete for P if every other language in P is log-space reducible to L.

We will see some examples in later lectures ...

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### Remark: Log-space Reductions for Larger Classes?

Could we use log-space reductions instead of polynomial reductions for defining hardness for other classes, e.g., for NP?

- Some authors do this (prominently Papadimitriou)
- All concrete polynomial reductions we have seen can be computed in logarithmic space

**Obvious question:** Are the classes "NP-complete problems under polynomial time reductions" and "NP-complete problems under log-space reductions" different?

Today's answer: Nobody knows (YCTBF)

(at least we have not seen any example of such differences, so it might not matter much in practice)

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### An NL-Complete Problem

#### Theorem 11.8: REACHABILITY is NL-complete.

**Proof idea:** We already showed membership. What remains is hardness.

Let  $\mathcal M$  be a non-deterministic log-space TM deciding  $\mathbf L$ .

On input w:

- (1) modify Turing machine to have a unique accepting configuration (easy)
- (2) construct the configuration graph (graph whose nodes are configurations of  $\mathcal{M}$  and edges represent possible computational steps of  $\mathcal{M}$  on w)
- (3) find a path from the start configuration to the accepting configuration

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### **NL-Completeness**

**Proof sketch:** We construct  $\langle G, s, t \rangle$  from  $\mathcal{M}$  and w using a log-space transducer:

- (1) A configuration  $(q, w_2, (p_1, p_2))$  of  $\mathcal{M}$  can be described in  $c \log n$  space for some constant c and n = |w|.
- (2) List the nodes of G by going through all strings of length  $c \log n$  and outputting those that correspond to legal configurations.
- (3) List the edges of G by going through all pairs of strings  $(C_1, C_2)$  of length  $c \log n$  and outputting those pairs where  $C_1 \vdash_{\mathcal{M}} C_2$ .
- (4) s is the starting configuration of G.
- (5) Assume w.l.o.g. that  $\mathcal{M}$  has a single accepting configuration t.

$$w \in \mathbf{L} \text{ iff } \langle G, s, t \rangle \in \mathbf{Reachability}$$

(see also Sipser, Theorem 8.25)

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#### coNL

As for time, we consider complement classes for space.

#### Recall Definition 9.6:

For a complexity class C, we define  $coC := \{L \mid \overline{L} \in C\}$ .

#### Complement classes for space:

- $coNL := \{ L \mid \overline{L} \in NL \}$
- coNPSpace :=  $\{L \mid \overline{L} \in NPSpace\}$

#### From Savitch's theorem:

PSpace = NPSpace and hence coNPSpace = PSpace, but merely NL  $\subseteq$  DSpace ( $\log^2 n$ ) and hence coNL  $\subseteq$  DSpace ( $\log^2 n$ )

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#### The NL vs. coNL Problem

Another famous problem in complexity theory: is NL = coNL?

- First stated in 1964 [Kuroda]
- Related question: are complements of context-sensitive languages also context-sensitive?
   (such languages are recognized by linear-space bounded TMs)
- Open for decades, although most experts believe NL ≠ coNL

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### The Immerman-Szelepcsényi Theorem

Surprisingly, two independent people resolve the NL vs. coNL problem simutaneously in 1987  $\,$ 

More surprisingly, they show the opposite of what everyone expected:

Theorem 11.9 (Immerman 1987/Szelepcsényi 1987): NL = coNL.

**Proof:** Show that **REACHABILITY** is in NL. (Why does this suffice?)

Remark: alternative explanations provided by

- Sipser (Theorem 8.27)
- Dick Lipton's blog entry We All Guessed Wrong (link)
- Wikipedia Immerman–Szelepcsényi theorem

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### Towards Nondeterminsitic Nonreachability

How could we check in logarithmic space that *t* is not reachable from *s*?

Initial idea: iterate through all reachable nodes looking for t

```
01 NaiveNonReach(G, s, t):
02 for each vertex v of G:
03 if CanReach(G, s, v) and v = t:
04 return FALSE
05 // eventually, if FALSE was not returned above:
06 return TRUE
```

Does this work?

No: the check CanReach(G, s, v) may fail even if v is reachable from s. Hence there are many (nondeterministic) runs where the algorithm accepts, although t is reachable from s.

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# Towards Nondeterminsitic Nonreachability

```
Things would be different if we knew
the number count of vertices reachable from s:
01 COUNTINGNONREACH(G, s, t, count):
02
    reached := 0
03
    for each vertex v of G:
       if CanReach(G, s, v):
04
05
         reached := reached + 1
         if v = t:
06
07
            return FALSE
     // eventually, if FALSE was not returned above:
80
     return (count = reached)
09
Problem: how can we know count?
```

### Counting Reachable Vertices – Intuition

#### Idea:

- Count number of vertices reachable in at most length steps
  - we call this number count<sub>length</sub>
  - then the number we are looking for is  $count = count_{|V(G)|-1}$
- Use a limited-length reachability test:
   CanReach(G, s, v, length): "t reachable from s in G in ≤ length steps"
   (we actually implemented CanReach(G, s, v) as CanReach(G, s, v, |V(G)| 1))
- Compute the count iteratively, starting with *length* = 0 steps:
  - for length > 0, go through all vertices u of G and check if they are reachable
  - to do this, for each such u, go through all v reachable by a shorter path, and check if you can directly reach u from them
  - use the counting trick to make sure you don't miss any v
     (the required number count<sub>length</sub> was computed before)

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### Counting Reachable Vertices – Algorithm

The count for length = 0 is 1. For length > 0, we compute as follows:

```
01 COUNTREACHABLE(G, s, length, count_{length-1}):
02
     count := 1 // we always count s
03
     for each vertex u of G such that u \neq s:
04
        reached := 0
05
       for each vertex v of G:
06
          if CanReach(G, s, v, length - 1):
            reached := reached + 1
07
80
            if G has an edge v \rightarrow u:
09
               count := count + 1
10
               GOTO 03 // continue with next u
        if reached < count_{length-1}:
11
12
          REJECT // whole algorithm fails
13
     return count
```

# Completing the Proof of NL = coNL

#### Putting the ingredients together:

```
01 NonReachable(G, s, t):
02 count := 1 // number of nodes reachable in 0 steps
03 for \ell := 1 to |V(G)| - 1:
04 count_{prev} := count
05 count := CountReachable(G, s, \ell, count_{prev})
06 return CountingNonReach(G, s, t, count)
```

It is not hard to see that this procedure runs in logarithmic space, since we use a fixed number of counters and pointers.

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# Summary and Outlook

Winning board games that don't allow moves to be undone is often PSpace-complete

L is the class of problems solvable using only a fixed number of linearly bound counters and pointers to the input

NL is the corresponding non-deterministic class, but we do not know if L = NL

### Summary:

#### What's next?

- So many ⊆! Will we ever get a strict ⊂?
- More generally: can more resources solve more problems?