



**Aalto University**  
School of Science

# CS-E4530 Computational Complexity Theory

Lecture 6: Complexity Classes and their Relationships

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# Agenda

- Basic requirements for complexity classes
- Time and space complexity classes
- Hierarchy theorems
- Reachability method
- Class inclusions
- Simulating nondeterministic space
- Closure under complement

(C. Papadimitriou: *Computational Complexity*, Chapter 7)

# 1. Basic Requirements for Complexity Classes

A complexity class is specified by

- model of computation (multi-string TMs)
- mode of computation (deterministic, nondeterministic, . . .)
- resource (time, space, . . .)
- bound (function  $f$ )

A *complexity class* is the set of all *languages* decided by some multi-string Turing machine  $M$  operating in the appropriate mode, and such that, for any input  $x$ ,  $M$  expends at most  $f(|x|)$  units of the specified resource.

## Reasonable Bound Functions

Not all functions provide reasonable bounds (are too complicated to compute within the bounds that they allow).

### Definition

A function  $f : \mathbf{N} \rightarrow \mathbf{N}$  is a *proper complexity function* if  $f$  is nondecreasing and there is a  $k$ -string TM  $M_f$  with input and output such that on any input  $x$ ,

1.  $M_f(x) = \sqsupset^{f(|x|)}$  where  $\sqsupset$  is a *tally* (“quasi-blank”) symbol,
2.  $M_f$  halts within  $O(|x| + f(|x|))$  steps, and
3.  $M_f$  uses  $O(f(|x|))$  space besides its input.

- Examples of proper complexity functions  $f(n)$ :

$$c, n, \lceil \log n \rceil, \log^2 n, n \log n, n^2, n^3 + 3n, 2^n, \sqrt{n}, n!, \dots$$

- If  $f$  and  $g$  are proper, so are, e.g.,  $f + g, f \cdot g, 2^g$ .
- Only proper complexity functions will be used as bounds.

## Precise Turing Machines

### Definition

Let  $M$  be a deterministic/nondeterministic multi-string Turing machine (with or without input and output).

Machine  $M$  is *precise* if there are functions  $f$  and  $g$  such that for every  $n \geq 0$ , for every input  $x$  of length  $n$ , and for every computation of  $M$ ,

1.  $M$  halts in precisely  $f(|x|)$  steps and
2. all of its strings (except those reserved for input and output whenever present) are at halting of length precisely  $g(|x|)$ .

(Precise bounds will be convenient in various simulation results).

## Simulating TMs with Precise TMs

### Proposition

*Let  $M$  be a deterministic or nondeterministic TM deciding a language  $L$  within time/space  $f(n)$  where  $f$  is proper.*

*Then there is a precise TM  $M'$  which decides  $L$  in time/space  $O(f(n))$ .*

### Proof sketch

The simulating machine  $M'$  on input  $x$

1. computes a yardstick/alarm clock  $\Gamma^{f(|x|)}$  using  $M_f$  (using a new set of strings) and
2. using the yardstick (output string of  $M_f$ )  
simulates  $M$  for exactly  $f(|x|)$  steps *or*  
simulates  $M$  using exactly  $f(|x|)$  units of space.

## 2. Time and Space Complexity Classes

- Given a proper complexity function  $f$ , we obtain following classes:  
**TIME**( $f$ ) (deterministic time)  
**NTIME**( $f$ ) (nondeterministic time)  
**SPACE**( $f$ ) (deterministic space)  
**NSPACE**( $f$ ) (nondeterministic space)
- The bound  $f$  can be a family of functions parameterized by a non-negative integer  $k$ ; meaning the union of all individual classes.

The most important are:  $\mathbf{TIME}(n^k) = \bigcup_{j>0} \mathbf{TIME}(n^j)$   
 $\mathbf{NTIME}(n^k) = \bigcup_{j>0} \mathbf{NTIME}(n^j)$

### Key Complexity Classes

<b>L</b>	=	<b>SPACE</b> ( $\log(n)$ )	<b>NL</b>	=	<b>NSPACE</b> ( $\log(n)$ )
<b>P</b>	=	<b>TIME</b> ( $n^k$ )	<b>NP</b>	=	<b>NTIME</b> ( $n^k$ )
<b>PSPACE</b>	=	<b>SPACE</b> ( $n^k$ )	<b>NPSPACE</b>	=	<b>NSPACE</b> ( $n^k$ )
<b>EXP</b>	=	<b>TIME</b> ( $2^{n^k}$ )			

The relationships of these classes will be studied in the sequel.

## Complements of Decision Problems

- Given an alphabet  $\Sigma$  and a language  $L \subseteq \Sigma^*$ , the *complement* of  $L$

$$\bar{L} = \Sigma^* - L.$$

- For a decision problem  $A$ , the answer for the complement “ $A$  COMPLEMENT” is “yes” iff the answer for  $A$  is “no”.

### Example

SAT COMPLEMENT: given a Boolean expression  $\phi$  in CNF, is  $\phi$  unsatisfiable?

### Example

REACHABILITY COMPLEMENT: given a graph  $(V, E)$  and vertices  $v, u \in V$ , is it the case that there is no path from  $v$  to  $u$ ?



## Closure under Complement

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

$$\{\bar{L} \mid L \in C\}.$$

**Example.** As  $\text{SAT} \in \text{NP}$ , then  $\text{SAT COMPLEMENT} \in \text{coNP}$ .

- All deterministic time and space complexity classes  $C$  are *closed under complement*: if  $L \in C$ , then  $\bar{L} \in C$ .

Proof. Exchange “yes” and “no” states of the deciding Turing machine.

- Hence, for example,  $\text{P} = \text{coP}$ .
- The same holds for nondeterministic *space* complexity classes (to be shown in the sequel).
- An important open question: are *nondeterministic time* complexity classes *closed under complement*?  
For instance, does  $\text{NP} = \text{coNP}$  hold?

### 3. Hierarchy Theorems

- We derive a quantitative hierarchy result:  
with sufficiently greater time allocation, Turing machines are able to perform more complex computational tasks.
- For a proper complexity function  $f(n) \geq n$ , define

$$H_f = \{M; x \mid M \text{ accepts input } x \text{ within } f(|x|) \text{ steps}\}.$$

- Thus,  $H_f$  is the time-bounded version of  $H$ , i.e. the language of the HALTING problem.

## Upper bound for $H_f$

### Lemma

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

### Proof sketch

A 4-string machine  $U_f$  deciding  $H_f$  in time  $f(n)^3$  is based on

- (i) the universal Turing machine  $U$ ,
- (ii) the single-string simulator of a multi-string machine,
- (iii) the linear speedup machine, and
- (iv) the machine  $M_f$  computing the yardstick of length  $f(n)$  where  $n$  is the length of the input  $x$ .

## Proof continued...

The machine  $U_f$  operates on input  $M;x$  as follows:

1.  $M_f$  computes the alarm clock  $\sqcap^{f(|x|)}$  for  $M$  (string 4).
2. The description of  $M$  is copied on string 3, string 2 is initialised to encode the initial state  $s$ , and string 1 the input  $\triangleright x$ .
3. Then  $U_f$  simulates  $M$  and advances the alarm clock. If  $U_f$  finds out that  $M$  accepts input  $x$  within  $f(|x|)$  steps, then  $U_f$  accepts, but if the alarm clock expires, then  $U_f$  rejects.

Observations:

- Since the multi-string machine  $M$  is simulated using a single string, each simulation step takes  $O(f(n)^2)$  time.
- The total running time is  $O(f(n)^3)$  for  $f(|x|)$  steps.
- The running time can be made at most  $f(n)^3$  by treating several symbols as one as in the proof of the linear speedup theorem.

## Lower bound for $H_f$

### Lemma

$$H_f \notin \mathbf{TIME}(f(\lfloor \frac{n}{2} \rfloor))$$

### Proof sketch

- Suppose there is a TM  $M_{H_f}$  that decides  $H_f$  in time  $f(\lfloor \frac{n}{2} \rfloor)$ .
- Consider  $D_f(M)$ : **if**  $M_{H_f}(M; M) = \text{“yes”}$  **then** “no” **else** “yes”.  
Thus,  $D_f$  on input  $M$  runs in time  $f(\lfloor \frac{2|M|+1}{2} \rfloor) = f(|M|)$ .
- If  $D_f(D_f) = \text{“yes”}$ , then  $D_f$  accepts input  $D_f$  within  $f(|D_f|)$  steps, hence  $D_f; D_f \in H_f$  and so  $M_{H_f}(D_f; D_f) = \text{“yes”}$ . But then  $D_f(D_f) = \text{“no”}$ , a contradiction.
- If  $D_f(D_f) = \text{“no”}$ , then  $D_f$  rejects input  $D_f$  within  $f(|D_f|)$  steps, hence  $D_f; D_f \notin H_f$  and so  $M_{H_f}(D_f; D_f) = \text{“no”}$ . But then  $D_f(D_f) = \text{“yes”}$ , a contradiction again.

## The time hierarchy theorem.

### Theorem

If  $f(n) \geq n$  is a proper complexity function, then the class  $\mathbf{TIME}(f(n))$  is strictly contained within  $\mathbf{TIME}((f(2n+1))^3)$ .

### Proof

- $\mathbf{TIME}(f(n)) \subseteq \mathbf{TIME}((f(2n+1))^3)$  as  $f$  is nondecreasing.
- By the first lemma:  $H_{f(2n+1)} \in \mathbf{TIME}((f(2n+1))^3)$ .
- By the second lemma:  
 $H_{f(2n+1)} \notin \mathbf{TIME}(f(\lfloor \frac{2n+1}{2} \rfloor)) = \mathbf{TIME}(f(n))$ .

### Corollary

$\mathbf{P}$  is a *proper* subset of  $\mathbf{EXP}$ .

### Proof

- Since  $n^k = O(2^n)$ , we have  $\mathbf{P} \subseteq \mathbf{TIME}(2^n) \subseteq \mathbf{EXP}$ .
- It follows by the time hierarchy theorem that  
 $\mathbf{TIME}(2^n) \subsetneq \mathbf{TIME}((2^{2n+1})^3) \subseteq \mathbf{TIME}(2^{2n^2}) \subseteq \mathbf{EXP}$ .

## The space hierarchy theorem

### Theorem

If  $f(n) \geq n$  is a proper complexity function, then the class  $\mathbf{SPACE}(f(n))$  is a *proper* subset of  $\mathbf{SPACE}(f(n) \log f(n))$ .

However, counter-intuitive results are obtained if non-proper complexity functions are allowed.

### Theorem (The Gap Theorem)

There is a recursive function  $f$  from the nonnegative integers to the nonnegative integers such that  $\mathbf{TIME}(f(n)) = \mathbf{TIME}(2^{f(n)})$ .

### Proof idea

The bound  $f$  can be defined so that no TM  $M$  computing on input  $x$  with  $|x| = n$  halts in any number of steps between  $f(n)$  and  $2^{f(n)}$ .

## 4. The Reachability Method

### Theorem

Let  $f(n)$  be a proper complexity function. Then

- (a)  $\text{SPACE}(f(n)) \subseteq \text{NSPACE}(f(n))$  and  $\text{TIME}(f(n)) \subseteq \text{NTIME}(f(n))$ .
- (b)  $\text{NTIME}(f(n)) \subseteq \text{SPACE}(f(n))$ .
- (c)  $\text{NSPACE}(f(n)) \subseteq \text{TIME}(c^{\log n + f(n)})$ .

### Proof

- (a) A TM is an NTM, too.
- (b) Simulation of all the choices within space  $f(n)$  (see below).
- (c) Proof by the reachability method (see below).



## Proof of $\text{NTIME}(f(n)) \subseteq \text{SPACE}(f(n))$

- Let  $L \in \text{NTIME}(f(n))$ . Hence, there is a precise nondeterministic Turing machine  $N$  that decides  $L$  in time  $f(n)$ .
- We show how to construct a deterministic machine  $M$  that simulates  $N$  within the space bound  $f(n)$ .
- Let  $d$  be the degree on nondeterminism of  $N$  (maximal number of possible moves for any state-symbol pair in  $\Delta$ ).
- Any computation of  $N$  on input  $x$  corresponds to an  $f(n)$ -long sequence of nondeterministic choices (represented by integers  $0, 1, \dots, d-1$ ) where  $n = |x|$ .
- The simulating deterministic machine  $M$  considers all such sequences of choices and simulates  $N$  on each.

## Proof—cont'd.

- With sequence  $(c_1, c_2, \dots, c_{f(n)})$   $M$  simulates the actions that  $N$  would have taken had  $N$  taken choice  $c_i$  at step  $i$ .
- If a sequence leads  $N$  to halting with “yes”, then  $M$  does, too. Otherwise it considers the next sequence. If all sequences are exhausted without accepting, then  $M$  rejects.
- There are an exponential number of simulations to be carried out, but they can all be completed in *space*  $f(n)$  by running them one-by-one, and always erasing the previous simulation to reuse space.
- As  $f(n)$  is proper, the first sequence  $0^{f(n)}$  can be generated in space  $f(n)$ .

## Proof of $\text{NSPACE}(f(n)) \subseteq \text{TIME}(c^{\log n + f(n)})$

The *reachability method* is used to prove the claim.

- Consider a  $k$ -string *nondeterministic* TM  $N$  with input and output which decides a language  $L$  within space  $f(n)$ .
- We develop a deterministic method for simulating the nondeterministic computation of  $N$  on input  $x$  within time  $c^{\log n + f(n)}$  where  $n = |x|$  and  $c$  is a constant depending on  $N$ .
- The *configuration graph*  $G(N, x)$  of  $N$  with input  $x$  is used: the vertices of the graph are all possible configurations of  $N$  with input  $x$  and there is an edge between two vertices (configurations)  $C_1$  and  $C_2$  iff  $C_1 \xrightarrow{N} C_2$ .
- Now  $x \in L$  iff there is a path from  $C_0 = (s, \triangleright, x, \triangleright, \epsilon, \dots, \triangleright, \epsilon)$  to some configuration of the form  $C = (\text{"yes"}, \dots)$  in  $G(N, x)$ .

## Proof—cont'd.

- A configuration  $(q, w_1, u_1, \dots, w_k, u_k)$  is a complete “snapshot” of a computation.
- Since  $N$  is a machine with input and output *deciding*  $L$  with the bound  $f(n)$ , we observe that for a configuration:
  - the output string can be neglected,
  - for the input string, only the cursor position can change, and
  - for all other  $k - 2$  strings, the length is at most  $f(n)$ .
- A configuration can be represented as  $(q, i, w_2, u_2, \dots, w_{k-1}, u_{k-1})$  where  $1 \leq i \leq n$  gives the cursor position on the input string.
- How many possible configurations does  $N$  have? At most

$$|K|(n+1)(|\Sigma|^{f(n)})^{2(k-2)} \leq |K|2n(|\Sigma|^{2(k-2)})^{f(n)} \leq nc_1^{f(n)} \leq c_1^{\log n + f(n)}$$

for some constant  $c_1$  depending on  $N$ .

## Proof—cont'd.

- Hence, deciding whether  $x \in L$  holds can be done by solving a reachability problem for a graph with at most  $c_1^{\log n + f(n)}$  vertices.
- The problem can be solved, say, with a quadratic algorithm in time  $c_2 c_1^{2(\log n + f(n))} \leq c^{\log n + f(n)}$  with  $c = c_2 c_1^2$ .
- The graph  $G(N, x)$  needs not to be represented explicitly (e.g., as an adjacency matrix) for the reachability algorithm.
- Given machine  $N$  the existence of an edge from  $C$  to  $C'$  can be determined on the fly by examining  $C$ ,  $C'$ , and the input  $x$ .

## 5. Class Inclusions

### Corollary

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

### Proof

1.  **$L = SPACE(\log n) \subseteq NSPACE(\log n) = NL$**  follows by (a).
2.  **$NL = NSPACE(\log n) \subseteq TIME(c^{\log n + \log n}) = TIME(n^{2 \log c}) \subseteq P$**  follows by (c).
3. By (a)  **$TIME(n^k) \subseteq NTIME(n^k)$**  which implies  **$P \subseteq NP$** .
4. By (b)  **$NTIME(n^k) \subseteq SPACE(n^k)$**  which implies  **$NP \subseteq PSPACE$** .
5. By (a) and (c)  **$SPACE(n^k) \subseteq NSPACE(n^k) \subseteq TIME(c^{\log n + n^k}) \subseteq TIME(2^{n^{k+c'}}) \subseteq EXP.$**

## Which inclusions are proper?

### Corollary

The class **L** is a proper subset of **PSPACE**.

### Proof

The space hierarchy theorem tells us  $\mathbf{L} = \mathbf{SPACE}(\log(n)) \subsetneq \mathbf{SPACE}(\log(n) \log(\log(n))) \subseteq \mathbf{SPACE}(n^2) \subseteq \mathbf{PSPACE}$ .

It is believed that *all* inclusions of the complexity classes in  $\mathbf{L} \subseteq \mathbf{NL} \subseteq \mathbf{P} \subseteq \mathbf{NP} \subseteq \mathbf{PSPACE} \subseteq \mathbf{EXP}$  are proper.

However, we only know that

- at least one of the inclusions between **L** and **PSPACE** is proper (but don't know which) and
- at least one of the inclusions between **P** and **EXP** is proper (but don't know which).

## 6. Simulating Nondeterministic Space

- The question is how efficiently can we simulate nondeterministic space by deterministic space?
- It follows by the previous theorem that

$$\mathbf{NSPACE}(f(n)) \subseteq \mathbf{TIME}(c^{\log n + f(n)}) \subseteq \mathbf{SPACE}(c^{\log n + f(n)}).$$

But can we do better than this?

- *Yes*, in fact, nondeterministic space can be simulated with quadratic deterministic space (using a theorem that follows).



## Savitch's theorem

### Theorem

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

### Proof sketch

- Given a graph  $G$  and vertices  $x, y$  and  $i \geq 0$ , define  $PATH(x, y, i)$ : there is a path from  $x$  to  $y$  of length at most  $2^i$ .
- If  $G$  has  $n$  vertices, any simple path is at most  $n$  long and we can solve reachability in  $G$  if we can compute whether  $PATH(x, y, \lceil \log n \rceil)$  holds for any given vertices  $x, y$  of  $G$ .
- This can be done using *middle-first search* within space bound  $\log^2 n$ .

## Proof continued

- **function**  $path(x, y, i)$  /\* middle-first search \*/  
**if**  $i = 0$  **then**  
    **if**  $x = y$  or there is an edge  $(x, y)$  in  $G$  **then** return “yes”  
**else for** all vertices  $z$  **do**  
    **if**  $path(x, z, i - 1)$  and  $path(z, y, i - 1)$  **then** return “yes”;  
    return “no”
- We prove that  $path(x, y, i)$  correctly determines  $PATH(x, y, i)$  by induction on  $i = 0, 1, 2, \dots$ :  
If  $i = 0$ , then clearly  $path$  correctly determines  $PATH(x, y, 0)$ .  
For  $i > 0$ ,  $path(x, y, i)$  returns “yes” iff there is a vertex  $z$  with  $path(x, z, i - 1)$  and  $path(z, y, i - 1)$  holding. By the inductive hypothesis there are paths from  $x$  to  $z$  and from  $z$  to  $y$  both at most  $2^{i-1}$  long. Then there is a path from  $x$  to  $y$  at most  $2^i$  long.

## Proof continued

- The algorithm is started with  $path(x, y, \lceil \log n \rceil)$ .
- $O(\log^2 n)$  space bound can be achieved by managing recursion using a stack containing a triple  $(x, y, i)$  for each active recursive call  $path(x, y, i)$  which is handled in the following way:
  - ▶ generate all vertices  $z$  one after the other reusing space.
  - ▶ For each vertex  $z$  put  $(x, z, i - 1)$  into the stack and call  $path(x, z, i - 1)$ .
  - ▶ If this returns “no”, then erase  $(x, z, i - 1)$  and move to the next vertex  $z$ .
  - ▶ If a positive answer is obtained, then erase  $(x, z, i - 1)$ , put  $(z, y, i - 1)$  into the stack and call  $path(z, y, i - 1)$ .
  - ▶ If this returns “no”, erase  $(z, y, i - 1)$  and move to the next vertex  $z$  otherwise return a positive answer to  $path(x, y, i)$ .
- As there are at most  $\log n$  recursive calls active with each taking at most  $3 \log n$  space, the  $O(\log^2 n)$  space bound is achieved.

## Corollary

For any proper complexity function  $f(n) \geq \log n$ ,

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

## Proof

- To simulate an  $f(n)$ -space bounded NTM  $N$  on input  $x$ , run the previous algorithm on the configuration graph  $G(N, x)$ .
- The edges of the graph  $G(N, x)$  are determined on the fly by examining the input  $x$ .
- The configuration graph has at most  $c_1^{\log n + f(n)} \leq c^{f(n)}$  vertices for some  $c$ .
- By Savitch's theorem, the algorithm needs at most  $(\log c^{f(n)})^2 = f(n)^2 \log^2 c = O(f(n)^2)$  space.

## Corollary

**PSPACE = NSPACE.**

 Nondeterminism is less powerful with respect to space than time.

## 7. Closure under Complement

- A key result about reachability will be established: the number of vertices reachable from a vertex  $x$  can be computed in nondeterministic  $\log n$  space!
- The complement (the number of vertices not reachable from  $x$ ) can be handled in nondeterministic  $\log n$  space, too! (This quantity can be obtained by a simple subtraction.)
- This implies that *nondeterministic space is closed under complement*.
- It is open (and doubtful) whether nondeterministic *time* complexity classes are closed under complement.

## Functions computed by NTMs

When does an NTM  $M$  compute a function  $F$  from strings to strings?

- On input  $x$ , each computation of  $M$  either
  - outputs the correct answer  $F(x)$  or
  - enters the rejecting “no” state.
- At least one computation must end up with  $F(x)$  which must be unique for all such computations.
- Such a machine observes a space bound  $f(n)$  iff for any input  $x$ , at halting all strings (except the ones reserved for input and output) are of length at most  $f(|x|)$ .

## Immerman-Szelepcsényi theorem

### Theorem

*Given a graph  $G$  and a vertex  $x$ , the number of vertices reachable from  $x$  in  $G$  can be computed by an NTM within space  $\log n$ .*

### Proof

- Let us define  $S(k)$  as the set of vertices in  $G$  which are reachable from  $x$  via paths of length  $k$  or less.
- The strategy is to compute values  $|S(1)|, |S(2)|, \dots, |S(n-1)|$  iteratively and recursively, i.e.  $|S(i)|$  is computed from  $|S(i-1)|$ .
- Given that the number of vertices in  $G$  is  $n$ , the number of vertices reachable from  $x$  in  $G$  is  $|S(n-1)|$ .
- Let  $G(v, u)$  mean that  $v = u$  or there is an edge from  $v$  to  $u$  in  $G$ .

## Proof continued

The nondeterministic algorithm:

```
|S(0)| ← 1;
for  $k \leftarrow 1, 2, \dots, n - 1$  do
   $l \leftarrow 0$ ;
  for each vertex  $u \leftarrow 1, 2, \dots, n$  do
    check whether  $u \in S(k)$  and set reply accordingly;
    /* See below how this is implemented */
    if reply = true then  $l \leftarrow l + 1$ ;
  end for;
  |S(k)| ←  $l$ 
end for
```



## Proof continued

```
/* Check whether  $u \in S(k)$  and set reply */  
 $m \leftarrow 0$ ;  $reply \leftarrow false$ ;  
for each vertex  $v \leftarrow 1, 2, \dots, n$  do  
  /* check whether  $v \in S(k-1)$  */  
   $w_0 \leftarrow x$ ;  $path \leftarrow true$   
  for  $p \leftarrow 1, 2, \dots, k-1$  do  
    guess a vertex  $w_p$ ; if not  $G(w_{p-1}, w_p)$  then  $path \leftarrow false$   
  end for  
  if  $path = true$  and  $w_{k-1} = v$  then  
     $m \leftarrow m + 1$ ; /*  $v \in S(k-1)$  holds */  
    if  $G(v, u)$  then  $reply \leftarrow true$   
  end if  
end for  
if  $m < |S(k-1)|$  then “give up” (end in “no” state)
```

## Proof continued

- Note that only  $\log n$ -space is needed as there are only nine variables:  $|S(k-1)|, k, l, u, m, v, p, w_p, w_{p-1}$  which each (an integer) can be stored in  $\log n$  space.
- The algorithm computes correctly  $|S(k)|$  (by induction on  $k$ ):
  - If  $k = 0$ , then  $|S(k)| = 1$  as given by the algorithm.
  - For  $k > 0$ , consider a computation that does not “give up”. We need to show that counter  $l$  is incremented iff  $u \in S(k)$ .  
If counter  $l$  is incremented, then  $reply = true$  implying that  $u \in S(k)$ , i.e. there is a path  $(x =)w_0, \dots, w_{k-1}(= v), u$ .  
If  $u \in S(k)$ , then there is some  $v \in S(k-1)$  such that  $G(v, u)$ . But as the computation does not “give up”,  $m = |S(k-1)|$  (which is the correct value by induction) and therefore all  $v \in S(k-1)$  are verified as such and, thus,  $reply$  is set to *true*.
  - Moreover, clearly there is at least one accepting computation where paths to the members of  $S(k-1)$  are correctly guessed.

## Closure under Complement

### Corollary

If  $f(n) \geq \log n$  is a proper complexity function, then  $\mathbf{NSPACE}(f(n)) = \mathbf{coNSPACE}(f(n))$ .

### Proof sketch

- Suppose  $L \in \mathbf{NSPACE}(f(n))$  is decided by an  $f(n)$ -space bounded NTM  $N$ . We build an  $f(n)$ -space bounded NTM  $\bar{N}$  deciding  $\bar{L}$ .
- On input  $x$ ,  $\bar{N}$  runs the previous algorithm on the configuration graph  $G(N, x)$  associated with  $N$  and  $x$ .
- $\bar{N}$  rejects if it finds an accepting configuration in any  $S(k)$ .
- Since  $G(N, x)$  has at most  $n_g = c^{f(n)}$  vertices, then  $\bar{N}$  can accept if  $|S(n_g - 1)|$  is computed without an accepting configuration.
- Due to bound  $n_g$ ,  $\bar{N}$  needs at most  $\log c^{f(n)} = O(f(n))$  space.

# Learning Objectives

- The definitions and background of major complexity classes: **P**, **NP**, **PSPACE**, **NPSpace**, **EXP**, **L**, and **NL**.
- The knowledge of basic relationships between complexity classes (inclusions and proper inclusions).
- Savitch's theorem and Immerman-Szelepcényi theorem.