

3rd Week



Space Complexity and the Linear Space Hypothesis

Synopsis.

- Circuit Complexity
- Non-Uniform Complexity Classes
- Parameterized Problems
- Sub-Linear-Space Computability
- Linear Space Hypothesis

April 23, 2018. 23:59

Course Schedule: 16 Weeks

Subject to Change

- **Week 1:** Basic Computation Models
- **Week 2:** NP-Completeness, Probabilistic and Counting Complexity Classes
- **Week 3:** Space Complexity and the Linear Space Hypothesis
- **Week 4:** Relativizations and Hierarchies
- **Week 5:** Structural Properties by Finite Automata
- **Week 6:** Type-2 Computability, Multi-Valued Functions, and State Complexity
- **Week 7:** Cryptographic Concepts for Finite Automata
- **Week 8:** Constraint Satisfaction Problems
- **Week 9:** Combinatorial Optimization Problems
- **Week 10:** Average-Case Complexity
- **Week 11:** Basics of Quantum Information
- **Week 12:** BQP, NQP, Quantum NP, and Quantum Finite Automata
- **Week 13:** Quantum State Complexity and Advice
- **Week 14:** Quantum Cryptographic Systems
- **Week 15:** Quantum Interactive Proofs
- **Week 16:** Final Evaluation Day (no lecture)

YouTube Videos

- This lecture series is based on numerous papers of **T. Yamakami**. He gave **conference talks (in English)** and **invited talks (in English)**, some of which were video-recorded and uploaded to YouTube.
- Use the following keywords to find a playlist of those videos.
- **YouTube search keywords:**
Tomoyuki Yamakami conference invited talk playlist



Conference talk video

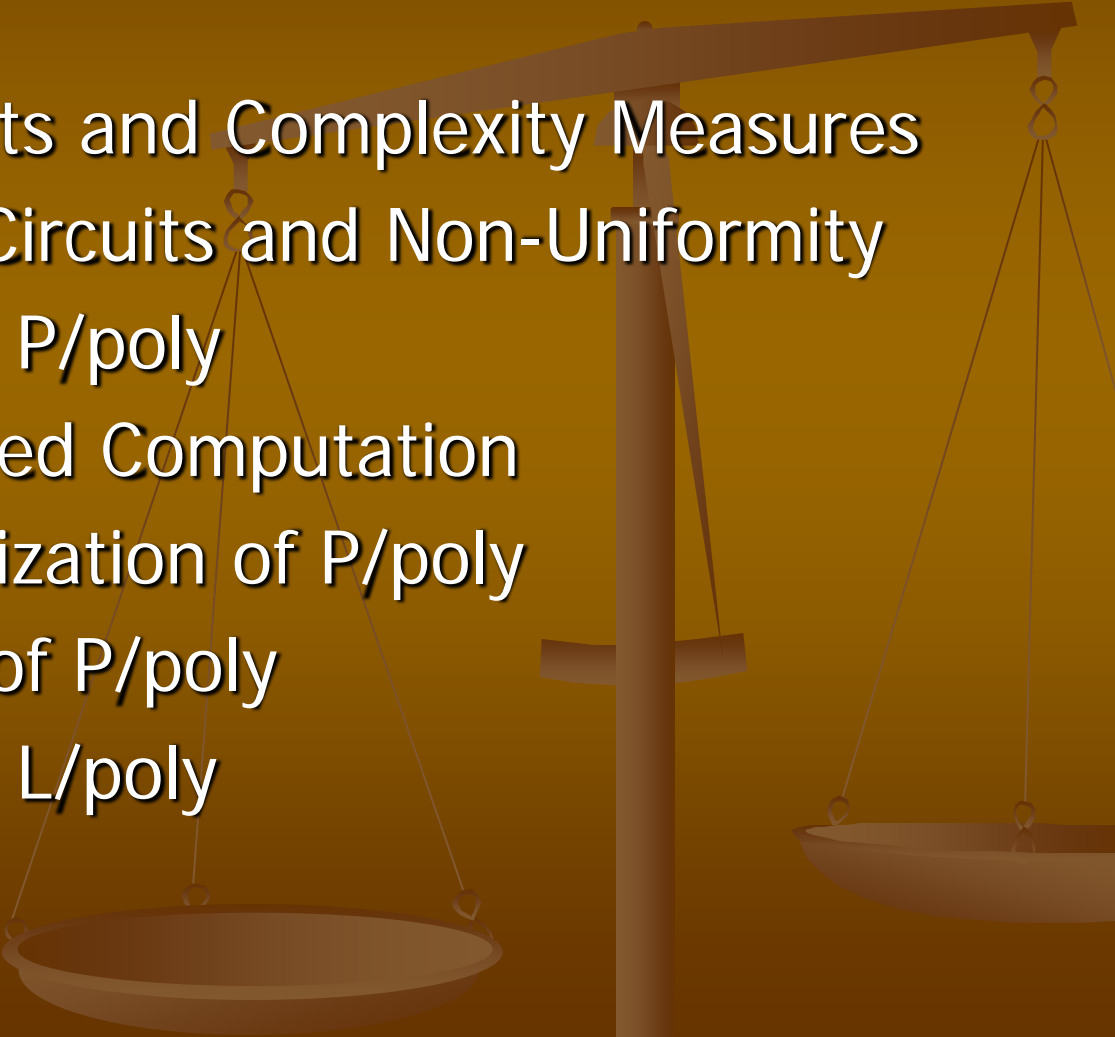


Main References by T. Yamakami



- ✎ **T. Yamakami.** Uniform-circuit and logarithmic-space approximations of refined combinatorial optimization problems. In Proc. of COCOA 2013, Lecture Notes in Computer Science, vol. 8287, pp. 318-329 (2013)
- ✎ **T. Yamakami.** The 2CNF Boolean formula satisfiability problem and the linear space hypothesis. In Proc. of MFCS 2017, LIPIcs 83, Schloss Dagstuhl - Leibniz-Zentrum fuer Informatik 62:1-62:14 (2017)
- ✎ **T. Yamakami.** Parameterized graph connectivity and polynomial-time sub-linear-space short reductions (preliminary report). In Proc. of RP 2017, Lecture Notes in Computer Science, vol. 10506, pp. 176-191 (2017)

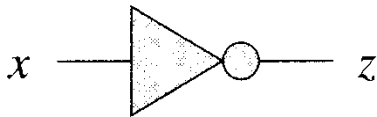
I. Non-Uniform Complexity Classes

1. Boolean Circuits
 2. Families of Circuits and Complexity Measures
 3. Polynomial-Size Circuits and Non-Uniformity
 4. Complexity Class P/poly
 5. Advice and Advised Computation
 6. Advice Characterization of P/poly
 7. Basic Properties of P/poly
 8. Complexity Class L/poly
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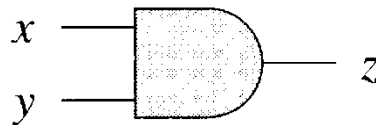
Boolean Circuits

- A Boolean circuit is composed of **logical gates** and **wires** (or edges) as illustrated below.

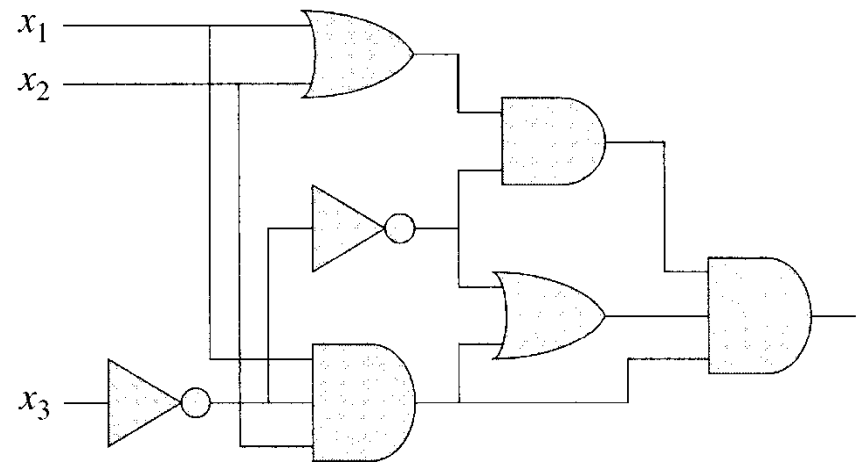
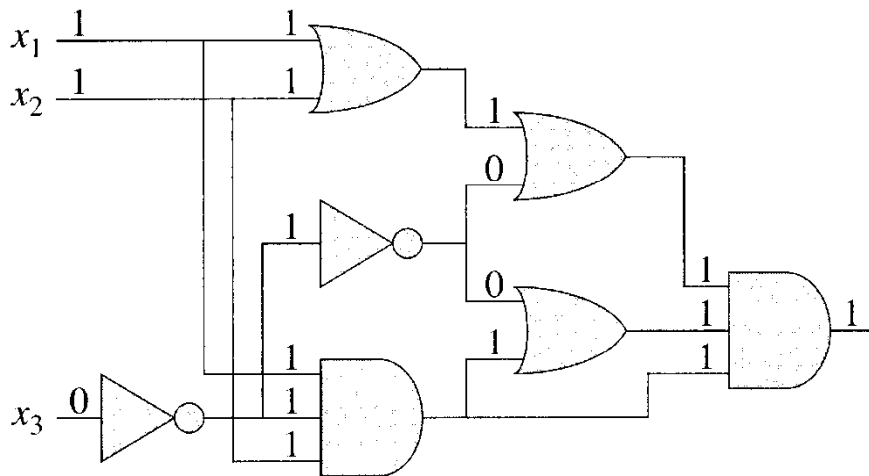
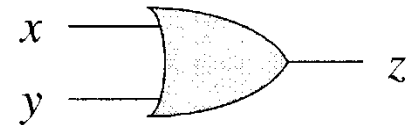
NOT gate



AND gate



OR gate



Truth Assignments of Boolean Circuits

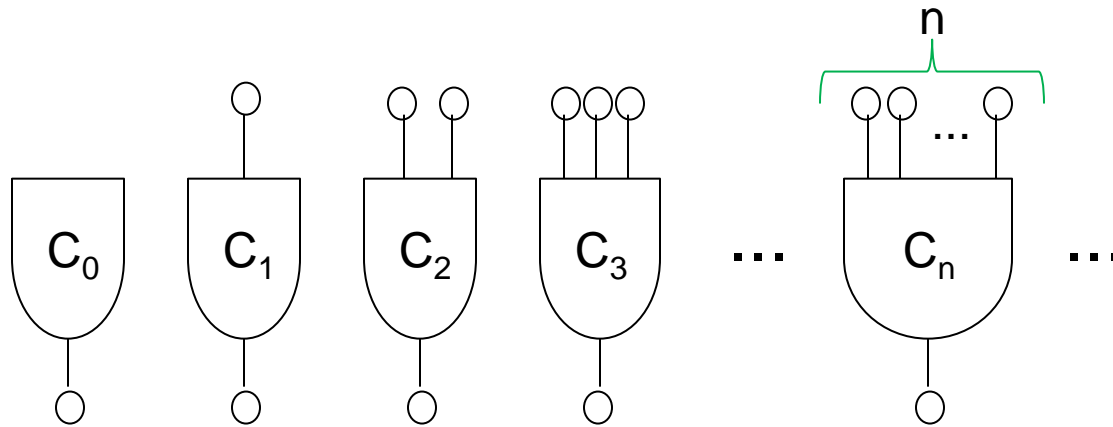
A truth assignment ($x_1 = 1, x_2 = 1, x_3 = 0$) causes the output to be 1.

Circuit-SAT

- A Boolean circuit C is called **satisfiable** if there exists a truth assignment by which C outputs 1.
- We consider the following problem, called Circuit-SAT.
- **Circuit Satisfiability Problem** (Circuit-SAT)
 - instance: a Boolean circuit C
 - question: is C satisfiable?
- **(Claim)** Circuit-SAT is NP-complete.
- Hence, $\text{Circuit-SAT} \in P \Leftrightarrow P = NP$.

Families of Circuits and Complexity Measures

- We consider a **family** $\{C_n\}_{n \in \mathbb{N}}$ of **Boolean circuits**, where each C_n denotes a Boolean circuit taking n -bit inputs.



We treat inputs and outputs as “gates” of indegree 0 and outdegree 0, respectively.

- We use the following complexity measures for circuits.

- **Circuit complexity measures:**

- **size** of circuit C = number of gates in C
- **depth** of circuit C = number of logical gates in the longest path from an input to an output

Polynomial-Size Circuits and Non-Uniformity

- We are interested in non-uniform families of Boolean circuits of polynomial size.
- A family $\{C_n\}_{n \in \mathbb{N}}$ of circuits **computes** (or **recognizes**) language L if, for any $n \in \mathbb{N}$ and for any Boolean values $x = x_1 x_2 \dots x_n \in \{0, 1\}^n$, $x \in L \Leftrightarrow C_{|x|}(x) = 1$.

- A family $\{C_n\}_{n \in \mathbb{N}}$ of circuits is said to be of **polynomial size** if there exists a nonnegative polynomial p such that, for any length n , C_n has size at most $p(n)$.

- We say that a family $\{C_n\}_{n \in \mathbb{N}}$ of circuits is **non-uniform** if there is no specific algorithm to produce a description (or an encoding) of C_n from input 1^n for every $n \in \mathbb{N}$.

Complexity Class P/poly

- **P/poly** is the collection of all decision problems (or languages) computed by non-uniform families of Boolean circuits of polynomial size.

More formally:

Such a family is called a non-uniform family of circuits

- For any decision problem L ,
 $L \in \text{P/poly} \Leftrightarrow$ there exist a constant $k \geq 1$ and a **non-uniform family** $\{ C_n \mid n \geq 1 \}$ **of Boolean circuits** such that
 - 1) $\text{Size}(C_n) = O(n^k)$ for every $n \geq 1$, and
 - 2) $x \in L \Leftrightarrow C_n(x) = 1$ for every x of length n .

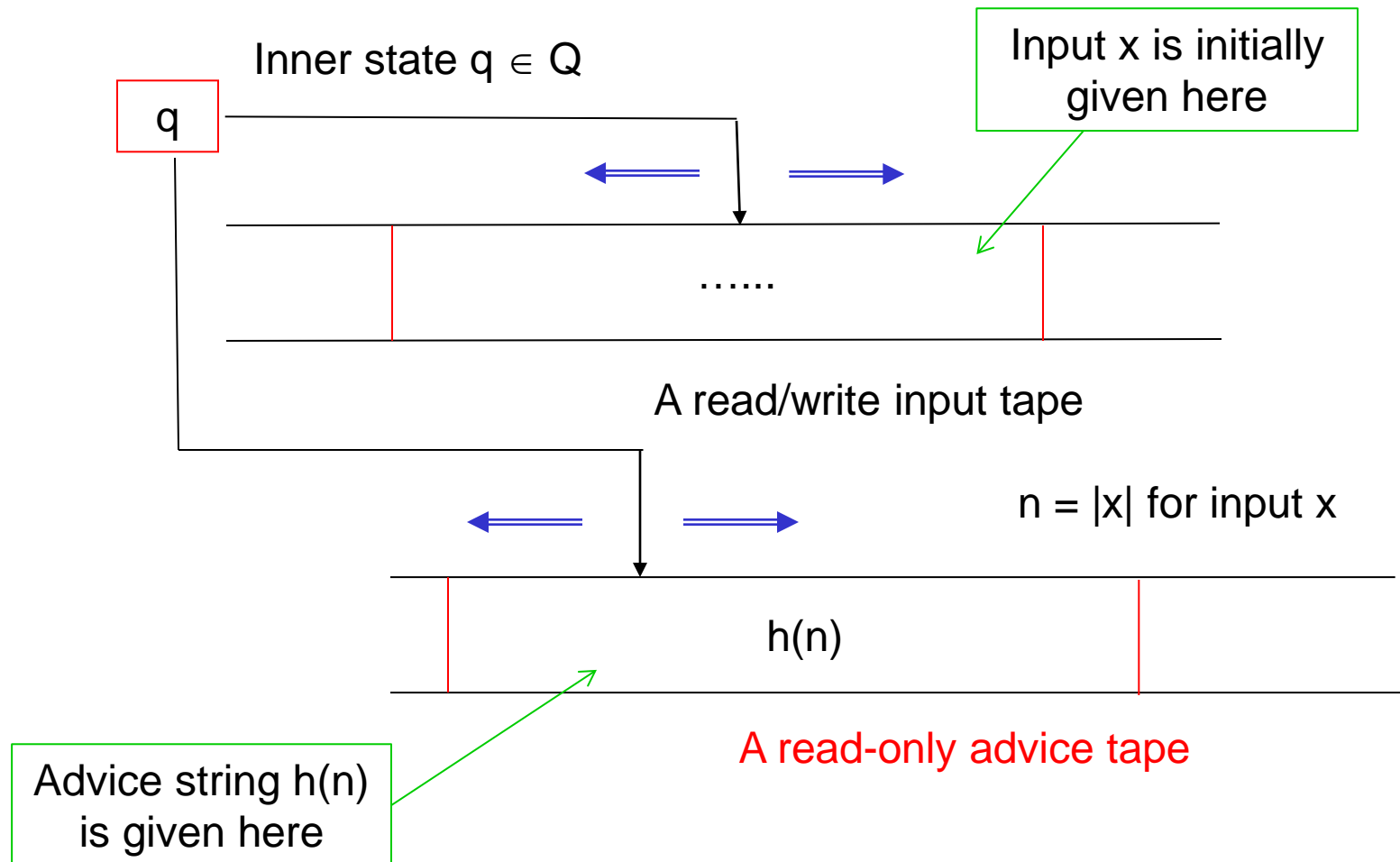
Advice and Advised Computation

- P/poly can be characterized in terms of advice.
- **Advice** is an external source that can provide with additional information to an underlying machine.
- Karp and Lipton (1982) considered the situation where a single **advice string** is given to underlying machines for each input length n .
- An **advice function** $h: \mathbb{N} \rightarrow \Sigma^*$ provides advice strings $h(n)$ for each input length n .
- An advised machine is a machine equipped with a read-only advice tape and it takes **two types of inputs**, a standard input string and also an advice string.

(See the next slide.)

Read-Only Advice Tapes

We provide a machine with an extra **read-only advice tape**.



Advice Characterization of P/poly I

- We give another characterization of P/poly using advised computation.
- We consider only advice of **polynomial length** (or size).
- For any decision problem L ,
 L is in **P/poly** \Leftrightarrow there exist a constant $k \geq 1$, a polynomial-time DTM M , an **advice function** $h: \mathbb{N} \rightarrow \Sigma^*$ for an advice alphabet such that
 - 1) $|h(n)| = O(n^k)$ for every input length n , and
 - 2) for every x , $x \in L \Leftrightarrow M$ accepts input pair $(x, h(|x|))$

This is expressed as $M(x, h(|x|)) = 1$

Advice Characterization of P/poly II

- In other words, for every language L ,

L is in $P/poly$ \Leftrightarrow there exist a language $A \in P$ and an advice function $h: N \rightarrow \Sigma^*$ such that, for every x ,

$$x \in L \Leftrightarrow \langle x, h(|x|) \rangle \in A.$$

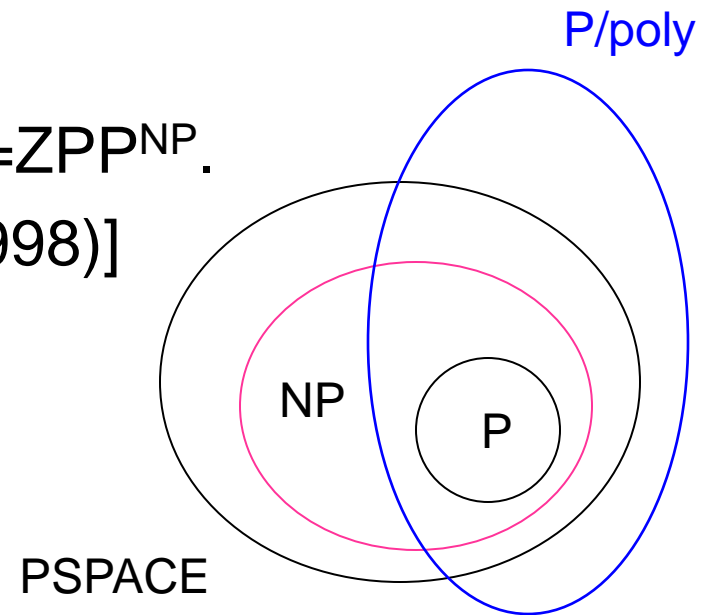
This is an encoding pair of $(x, h(|x|))$.

- By changing “P” in the above definition with other complexity classes C , we can define other advised complexity classes $C/poly$.
- For example, we obtain NP/poly, BPP/poly, UP/poly, etc.

(*) UP will be discussed in Week 4.

Basic Properties of P/poly

- Note that P/poly contains **non-recursive problems** (that is, problems that cannot be solved by any algorithm) because advice functions may not generally be computable.
- **(Claim)** $BPP \subseteq P/poly$
- **(Claim)** If $NP \subseteq P/poly$, then $PH = ZPP^{NP}$.
[Köbler-Watanabe (1998)]
- **Open Problems:**
 - Does $NP \subseteq P/poly$?
 - Does $PSPACE \subseteq P/poly$?



Many researchers believe in this way.

Complexity Class L/poly

- The use of advice gives rise to many non-uniform complexity classes.
- Here, we introduce another complexity class L/poly using **log-space DTMs** with polynomial-size advice.
- Let S be any decision problem or a language.

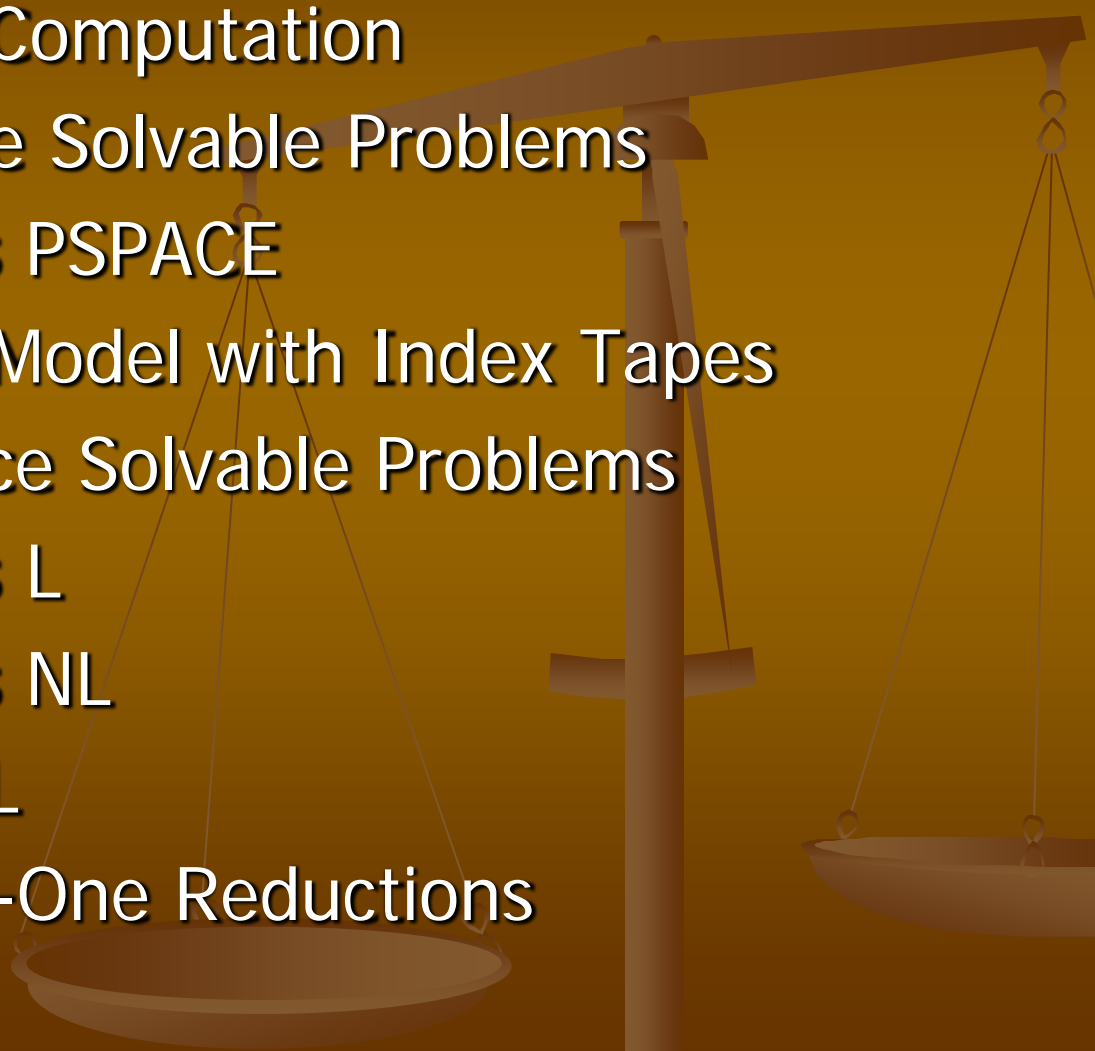
S is in **L/poly** \Leftrightarrow there exist a constant $k \geq 1$, a **log-space DTM** M , an advice function $h: \mathbb{N} \rightarrow \Sigma^*$ such that

- 1) $|h(n)| = O(n^k)$ for every input length n , and
- 2) for any x , $x \in S \Leftrightarrow M(x, h(|x|)) = 1$.

- **(Claim)** $L \subseteq L/poly$
- **Open Problem:** Is $NL \subseteq L/poly$?

Log-space DTMs will be explained shortly.

II. Space-Bounded Computation

1. Space-Bounded Computation
 2. Polynomial-Space Solvable Problems
 3. Complexity Class PSPACE
 4. Random Access Model with Index Tapes
 5. Logarithmic-Space Solvable Problems
 6. Complexity Class L
 7. Complexity Class NL
 8. Function Class FL
 9. Log-Space Many-One Reductions
- 

Space-Bounded Computation

- Earlier, we have discussed **time-bounded computation** and **time-bounded solvable problems**.
- Here, we are focused on space-bounded computation and associated problems.
- Let s be a space-bounding function from \mathbb{N} to \mathbb{N} such that **$s(n) \geq \log(n)$ for all $n \geq 1$** .
- We say that an algorithm (i.e., a DTM) **solves** a (decision) problem **using space $O(s(n))$** if, when it is provided a problem instance x of length n , the algorithm can produce the solution using $O(s(n))$ space.
- There is **no bound for running time** but the algorithm must halt eventually.

Polynomial-Space Solvable Problems

- A problem is said to be **$s(n)$ -space solvable** if there exists an algorithm to solve it using space $O(s(n))$.
- When $s(n)$ is a polynomial (i.e., $s(n)=O(n^k)$ for some constant k), a problem is said to be **polynomial-space solvable**.
- Note that any algorithm that runs in time $t(n)$ also uses space at most $t(n)$.
- Hence, polynomial-time solvable problems are also polynomial-space solvable.
- However, the converse does not hold in general.

How to Solve NP-Complete Problems

- Using polynomial-space, we can easily solve NP-complete problems.
- Take Circuit-SAT as an example of NP-complete problems.
- Consider this algorithm. \Rightarrow
- This algorithm uses only $O(n)$ bits to remember v and $O(|\text{code}(C)|^2)$ bits to simulate $C(v)$.
- Hence, Circuit-SAT is polynomial-space solvable.

Algorithm for Circuit-SAT

1. Take Boolean circuit C as an input.
2. Set $v=0^n$.
3. Check if $C(v) = 1$.
4. If so, accept and halt.
5. Else, if $v = 1^n$, reject and halt.
6. Else, increment v by one and go to Step 3.

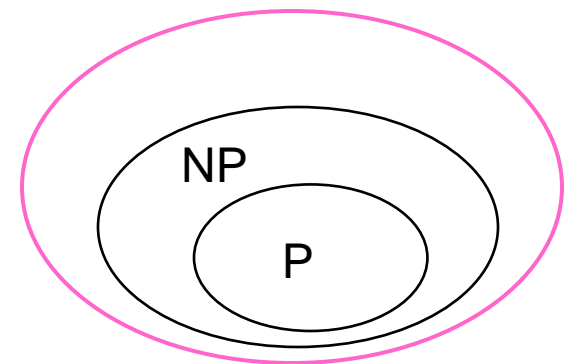
$v =$

00.....00	↓ increment
00.....01	
00.....10	
⋮	
11.....11	

Complexity Class PSPACE

- We introduce a complexity class defined by deterministic polynomial-space computations.
- A decision problem L is in **PSPACE** if there is a DTM M such that, for any input x ,
 1. $x \in L \rightarrow M$ accepts x ,
 2. $x \notin L \rightarrow M$ rejects x , and
 3. M uses polynomial space.

PSPACE = NPSPACE



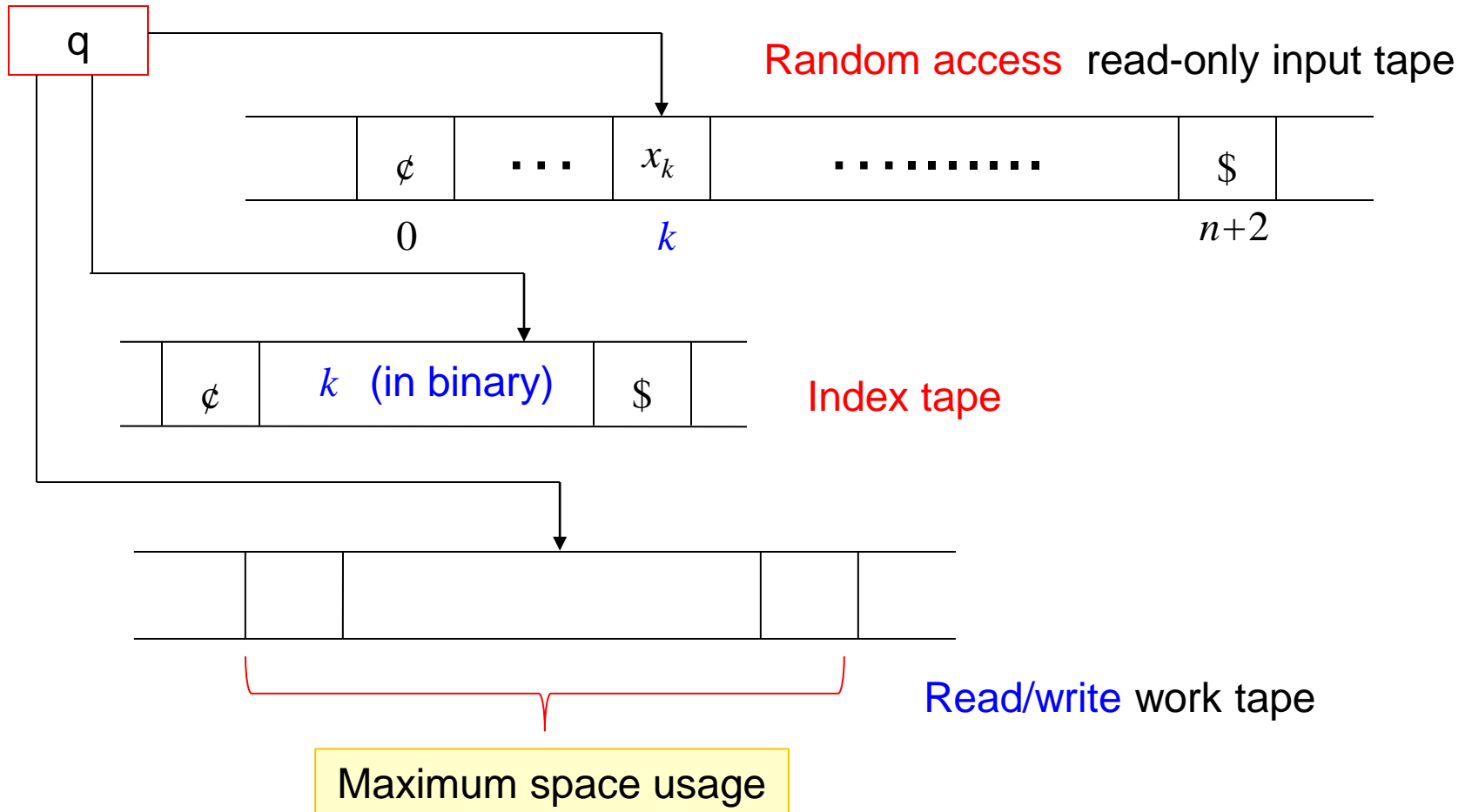
- **(Claim)** $P \subseteq NP \subseteq PSPACE \subseteq EXP$.
- **(Claim)** $PSPACE = NPSPACE$. [Savitch (1970)]

Function Class FPSPACE

- Next, we consider functions $f : \Sigma^* \rightarrow \Sigma^*$ (where Σ is an alphabet).
- A function $f : \Sigma^* \rightarrow \Sigma^*$ is in **FPSPACE** \Leftrightarrow
 1. f is p -bounded (i.e., $|f(x)| = O(|x|^k)$ for some $k \geq 1$), and
 2. there is a DTM M such that, for any input x , M produces $f(x)$ on the output tape using space $O(\log(|x|))$.
- **(Claim)** $PF \subseteq FPSPACE$.

Random Access Model with Index Tapes

- To consider logarithmic-space computation, we need a **random access model** of multi-tape Turing machines.



How to Operate a Machine

- To read each symbol written on an input tape, we need to take a series of steps described below.
 1. A machine M writes down an index k in binary on the index tape.
 2. M enters a special state, called an index state q_{index} , to initiate the process of random accessing.
 3. An input-tape head of M jumps to the cell indexed k .
 4. M scans the k -th tape cell and then the index tape is automatically become empty.
- This process is repeatedly taken to read all or some input symbols.

Logarithmic-Space Solvable Problems

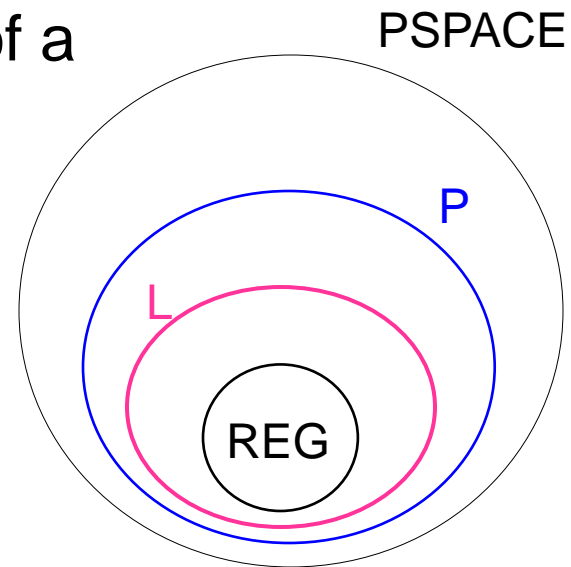
- Here, when we consider the space usage of a machine, **we do not include any read/write work tape.**
 - Let s be a function from \mathbb{N} to \mathbb{N} .
 - We say that an algorithm (or a deterministic Turing machine) **solves** a problem A **using space $O(s(n))$** if, for any instance x of length n , the algorithm can produce a solution of A using $O(s(n))$ space.
- A problem is **logarithmic-space (or log-space) solvable** if there exists an algorithm to solve it using $O(\log n)$ space.

Complexity Class L

- A decision problem (or a language) A is in L if there is a DTM M such that, for any input x ,
 1. $x \in A \rightarrow M$ accepts x ,
 2. $x \notin A \rightarrow M$ rejects x , and
 3. M uses logarithmic space (or log space).

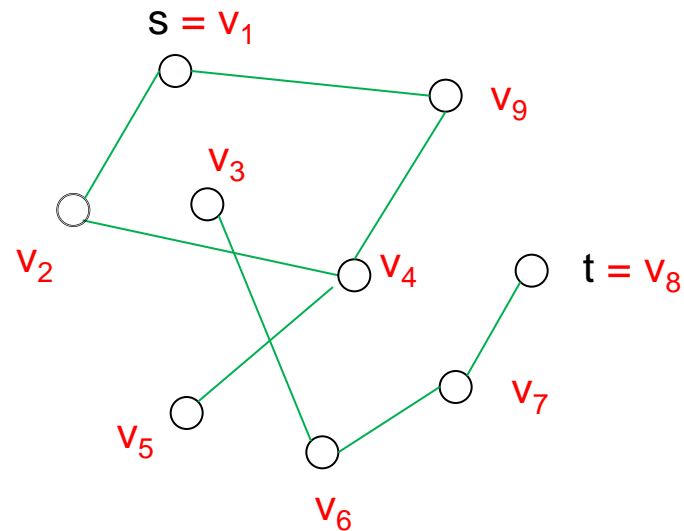
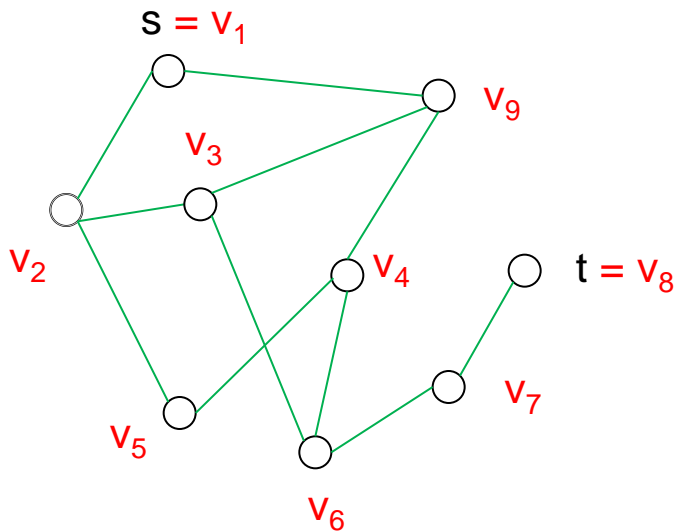
- It is possible to **trim** the running time of a machine to “**polynomial time.**”

- **(Claim)** $REG \subseteq L \subseteq P$.
- **(Claim)** $L \neq PSPACE$. [Savitch (1970)]



USTCON: Typical Problem in L

- Complexity class **L** contains the following problem.
- **Undirected s-t Connectivity Problem (USTCON)**
 - **instance:** an undirected graph G and two vertices s, t
 - **question:** Is there a path between s and t ?



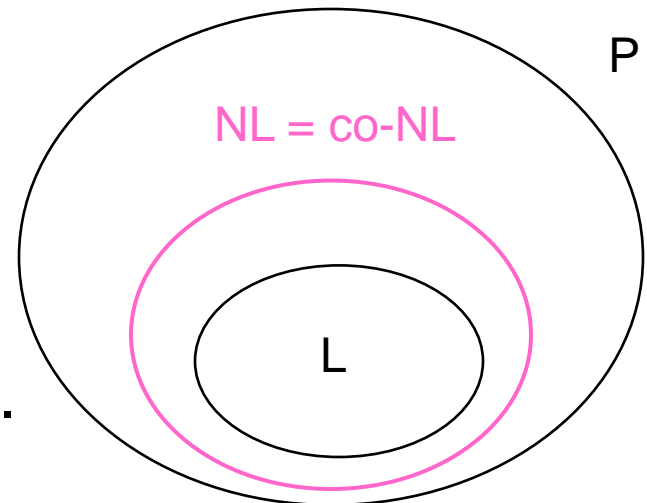
Complexity Class NL

- A decision problem (or a language) L is in **NL** if there is an **NTM** (nondeterministic Turing machine) M such that, for any input x ,
 1. $x \in L \leftrightarrow$ there exists an accepting computation path of M on x (or x is accepted by M), and
 2. M uses logarithmic space (or log space) on all inputs.

- **(Claim)** $L \subseteq NL \subseteq P$

- **(Claim)** $NL = \text{co-NL}$
[Immerman (1988),
Szelepcsényi (1988)]

- Hence, NL looks different from NP.



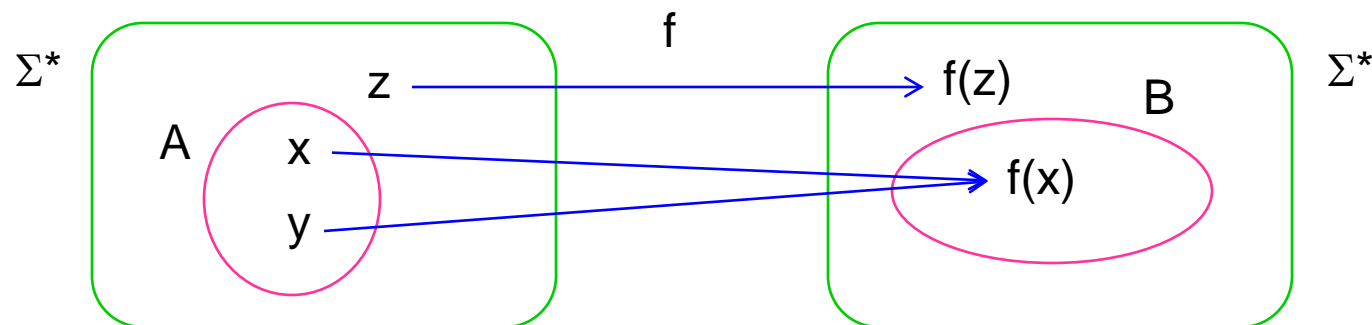
Function Class FL

- Let us recall the function class **FP** from Week 1.
- Here, we consider a log-space version of FP.
- Let $f: \Sigma^* \rightarrow \Sigma^*$ be any function, where Σ is an alphabet.
- This function $f: \Sigma^* \rightarrow \Sigma^*$ is called **log-space computable** if
 1. f is **p-bounded** (i.e., $|f(x)| = O(|x|^k)$ for some $k > 0$),
 2. there exists a DTM M with an output tape such that, on each input $x \in \Sigma^*$, M produces $f(x)$ on its output tape, and
 3. on input x , M uses only $O(\log(n))$ space on the work tape (but no space bound is imposed on the output tape).
- Let **FL** denote the collection of all p-bounded log-space computable functions.

Log-Space Many-One Reductions

- A language A is **log-space many-one reducible** (**L-m-reducible** or **\leq_m^L -reducible**) to language B if there exists a log-space DTM M such that, for any input x ,
 - $x \in A \Leftrightarrow M$ on input x produces y , which is p -bounded, and $y \in B$.
- In this case, we write **$A \leq_m^L B$** .
- In other words,

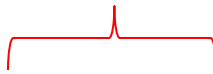
$$A \leq_m^L B \Leftrightarrow \exists f \in \text{FL} \forall x \in \Sigma^* [x \in A \Leftrightarrow f(x) \in B]$$



2SAT is NL-Complete

- Recall from Week 2 that **3SAT** is NP-complete.
- Complexity class **NL** contains the following problem.
- **2-Satisfiability Problem (2SAT)**
 - instance: a Boolean formula φ of 2CNF (2-conjunctive normal form)
 - question: Is φ satisfiable?
- E.g., 2CNF: $\varphi \equiv (x_1 \vee x_2) \wedge (x_1 \vee \neg x_3) \wedge (\neg x_2 \vee x_3)$

2 literals


- **(Claim)** 2SAT is NL-complete. [Jones (1975)]

$2SAT_k$ is also NL-Complete

- It turns out that 2SAT is not suitable for our purpose.
- Thus, we consider a restricted variant of 2SAT.
- $2SAT_k$ is the set of all 2SAT formula, each variable of which appears as literals at most k times.

- **Example:** $k=3$

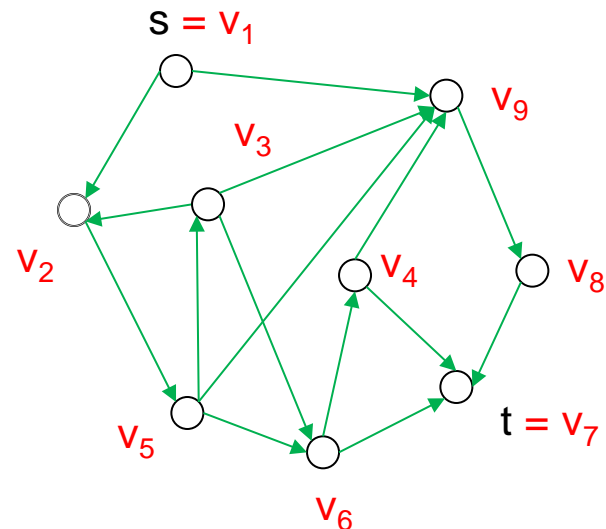
$$\varphi \equiv (x_1 \vee \neg x_6) \wedge (x_2 \vee x_3) \wedge (\neg x_5 \vee x_2) \wedge (\neg x_4 \vee \neg x_2)$$
$$m_{\text{vbl}}(\varphi) = 6, m_{\text{cls}}(\varphi) = 4$$

Each x_i appears at most 3 times

- **(Claim)** $2SAT_k$ ($k \geq 3$) is NL-complete.
- However, it is not known that $2SAT_k \notin L$.

DSTCON is NL-complete

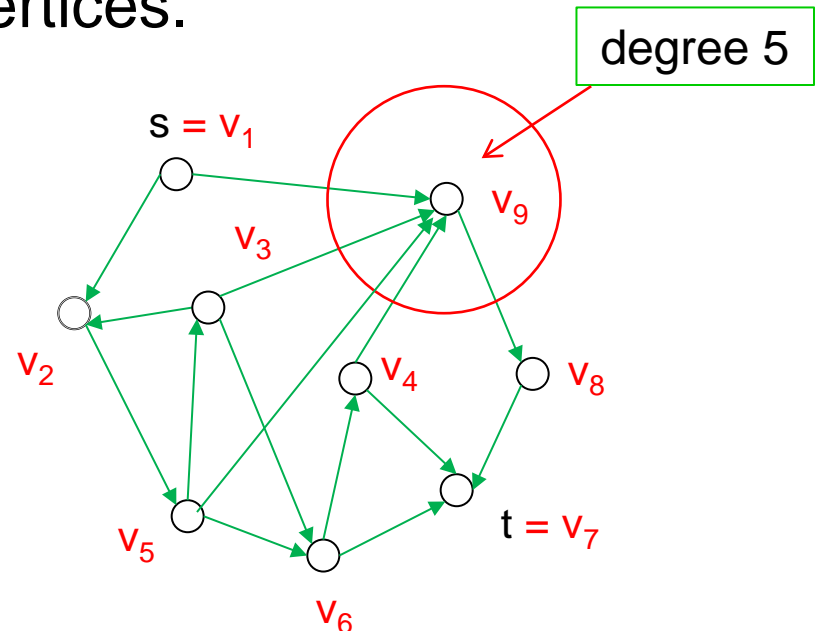
- Complexity class **NL** contains the following problems.
- **Directed s-t Connectivity Problem (DSTCON)**
 - **instance:** a directed graph G and two vertices s, t
 - **question:** Is there a path from s to t ?



- **(Claim)** DSTCON is NL-complete. [Jones (1975)]

kDSTCON is also NL-complete

- Consider a restricted variant of DSTCON.
- The **degree** of a vertex (or a node) is the number of edges connected to the vertex.
- **kDSTCON** consists of DSTCON instances whose graphs have **degree at most k** at all vertices.
- **(Claim)** For any constant $k \geq 3$, kDSTCON is NL-complete.
- However, it is not known that $3\text{DSTCON} \notin \text{L}$.

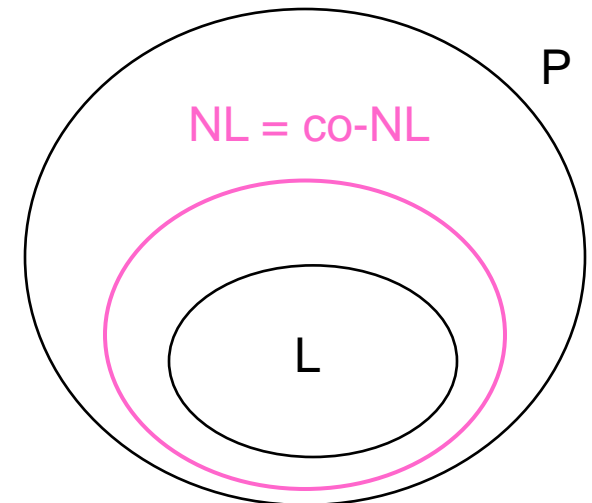
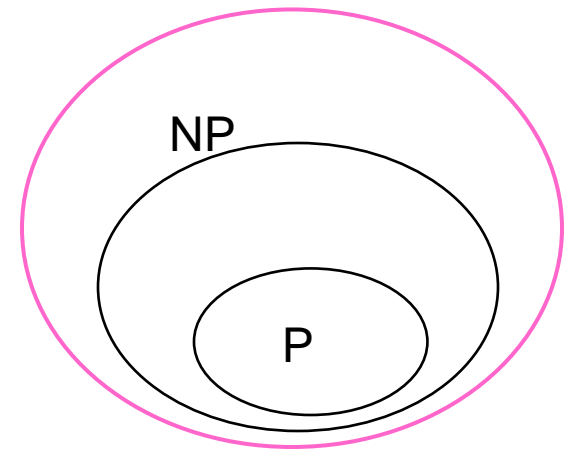


Open Problems

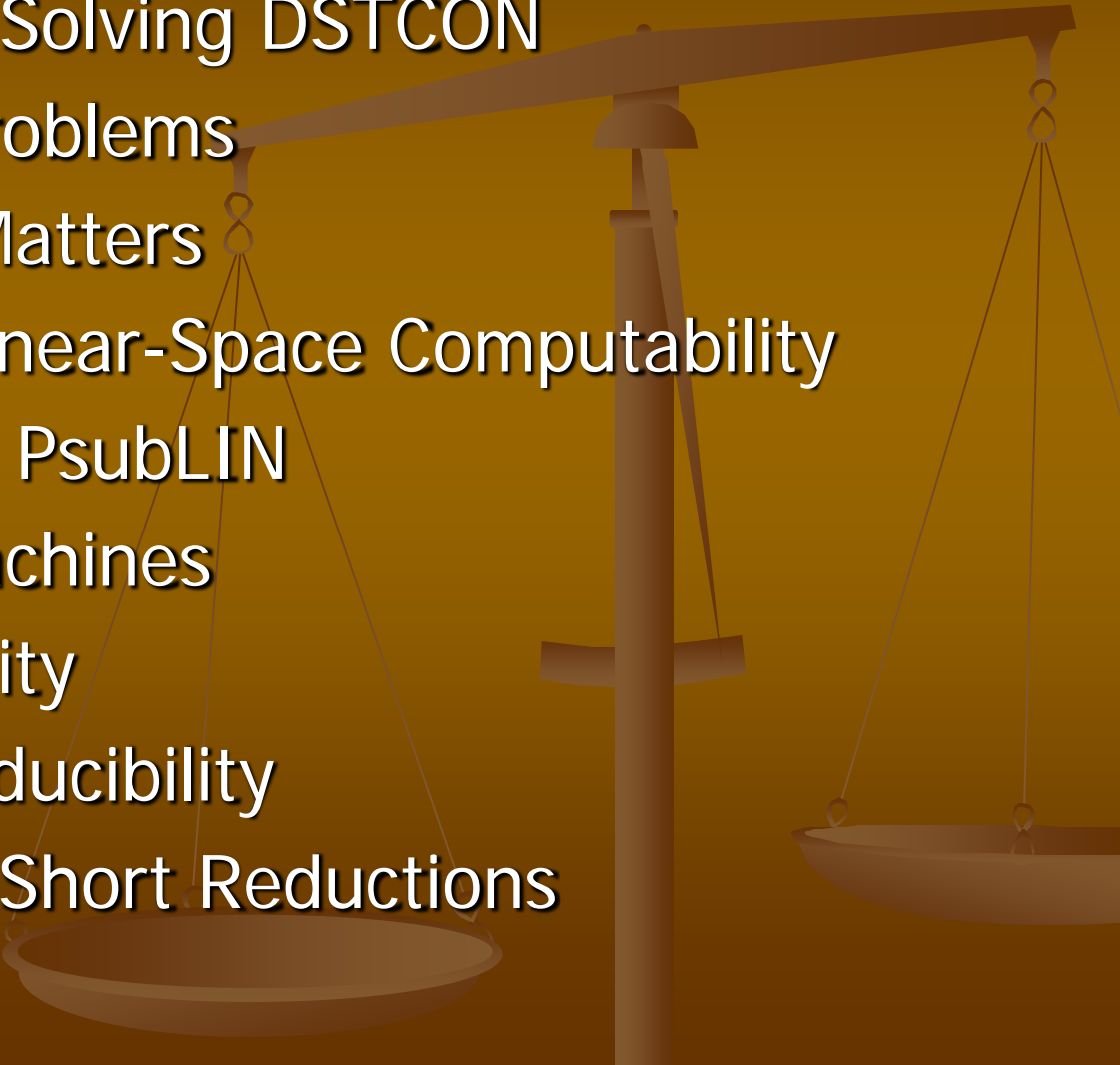
- The following questions regarding L, NL, and PSPACE are not yet answered.

- Is $P = PSPACE$?
- Is $NP = PSPACE$?
- Is $L = P$?
- Is $L = NL$?
- Is $NL = NP$?
- Is $CFL \subseteq L$?

PSPACE = NPSPACE



III. Sub-Linear-Space Computability

1. Space Usage for Solving DSTCON
 2. Parameterized Problems
 3. Size Parameter Matters
 4. Poly-Time Sub-Linear-Space Computability
 5. Complexity Class PsubLIN
 6. Oracle Turing Machines
 7. SLRF-T-Reducibility
 8. Short SLRF-T-Reducibility
 9. Relationships by Short Reductions
- 

Space Usage for Solving DSTCON

- Consider the following directed s-t connectivity problem.
- **DSTCON(m,n)**
 - **instance:** a directed graph G of **n vertices** and **m edges**, and two vertices s, t
 - **question:** is there any path from s to t?
- Barnes, Buss, Ruzzo, and Schieber (1998) gave an algorithm that solves DSTCON(m,n) in $O(m+n)$ time using $n^{1-c/\sqrt{\log(n)}}$ **space** for an appropriate constant $c>0$.
- **Open Problem:**
Can we improve the above space bound down to $O(n^\varepsilon \text{ polylog}(m+n))$ for certain $\varepsilon \in [0,1)$?

Size Parameters

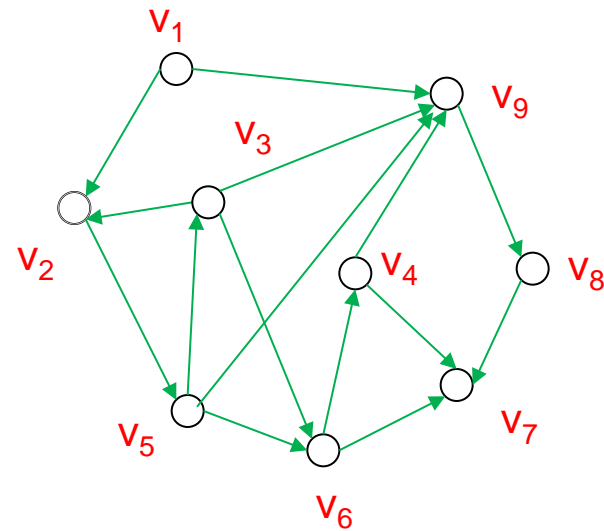
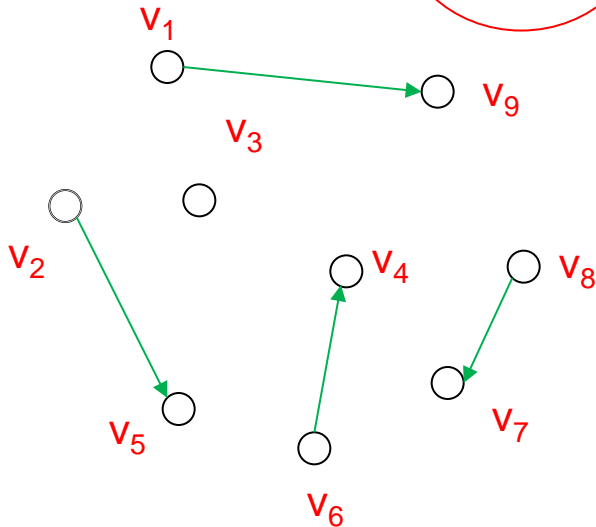
- It is useful to **parameterize** problems by taking appropriate size parameters.
- A **size parameter** $m: \Sigma^* \rightarrow \mathbb{N}$ is a function that gives a “size” $m(x)$ of an input x (e.g., $m(x) = |x|$).
- **Here are 2 simple examples.**
- For a CNF Boolean formula φ :
 - $m_{\text{vbl}}(\varphi)$ = number of different variables in φ
 - $m_{\text{cls}}(\varphi)$ = number of clauses in φ
- For a directed/undirected graph G :
 - $m_{\text{ver}}(G)$ = number of vertices in G
 - $m_{\text{edg}}(G)$ = number of edges in G

Size Parameter Matters

- How different is the gap between $m_{\text{ver}}(G)$ and $m_{\text{edg}}(G)$?
 - ✓ $m_{\text{ver}}(G)$ = number of vertices in G
 - ✓ $m_{\text{edg}}(G)$ = number of edges in G

$m_{\text{ver}}(G) \leq 2m_{\text{edg}}(G)$ if there is no isolated vertex

$m_{\text{edg}}(G) \leq m_{\text{ver}}(G)^2$ A huge difference!



Parameterized Problems

- In practice, execution time and space usage are often measured according to size $m(x)$ of input x .
- Impagliazzo, Paturi, and Zane (2001) took a new approach toward kSAT and Search-kSAT, parameterized by m_{vbl} and m_{cls} .
- A **parameterized decision problem** is a pair (A, m) of a (standard) decision problem $A \subseteq \Sigma^*$ and a size parameter $m: \Sigma^* \rightarrow \mathbb{N}$.
- Parameterized decision problem (A, m)
 - **instance**: x with size $m(x)$
 - **question**: is $x \in A$?

Poly-Time Sub-Linear-Space Computability

- We use deterministic Turing machines (DTMs), each of which has an input tape and a work tape.
- We are interested in DTMs that use only $O(n^c)$ time and restricted space to solve given decision problems.
- Let m be a size parameter.

- An informal term “sub linear w.r.t. m ” means

$$m(x)^\varepsilon \text{ polylog}(|x|)$$

for a fixed constant $\varepsilon \in [0, 1)$ and a polylogarithmic function $\text{polylog}(n)$ (i.e., $\text{clog}^k(n) + d$ for some $c > 0$ and $k \geq 0$).

- Here, we are focused on deterministic algorithms that run in polynomial time using only sub-linear space.

PTIME,SPACE(...)

- It is useful in practice to introduce a new notation.
- Let (L,m) denote a **parameterized problem**, where L is a decision problem and m is a size parameter.
- $(L,m) \in \text{PTIME,SPACE}(f(n)) \Leftrightarrow$
 $\exists M:\text{DTM s.t. } \forall x$
 - 1) $x \in L \rightarrow M$ accepts x
 - 2) $x \notin L \rightarrow M$ rejects x
 - 3) M runs in time polynomial in $|x|$ using $O(f(m(x)))$ space.

$O(|x|^k)$ time for some $k > 0$

Complexity of 2SAT(m,n)

- **Theorem:** [Yamakami (2017)]
 $\exists c > 0 \exists l: \text{polylog function s.t.}$
 $2\text{SAT}(m,n) \in \text{PTIME}, \text{SPACE}(n^{1-c/\sqrt{\log(n)}} l(m+n)),$
where a 2SAT(m,n)-instance has n variables and m clauses.
- The proof of the above theorem follows directly from Barnes, Buss, Ruzzo, and Schiebe's (1998) fast algorithm for DSTCON.
- **Open Problem:**
Is it true that $2\text{SAT}(m,n) \in \text{PTIME}, \text{SPACE}(n^\varepsilon)$ for a constant $\varepsilon \in (0, 1)$?

Complexity Class PsubLIN

- We define a new practical complexity class called **PsubLIN**.

- ✓ “P” stands for “**polynomial-time.**”

- ✓ “subLIN” stands for “**sub-linear space.**”

- **PsubLIN** = class of (parameterized) decision problems or search problems (L, m) such that L is solved in time polynomial in $|x|$ using sub-linear space (w.r.t. m)

- That is,

$$\mathbf{PsubLIN} = \cup_{0 \leq \varepsilon < 1} \mathbf{PTIME, SPACE}(m(x)^\varepsilon \text{polylog}(|x|))$$

- **(Claim)** $L \subseteq \mathbf{PsubLIN} \subseteq \mathbf{P}$.

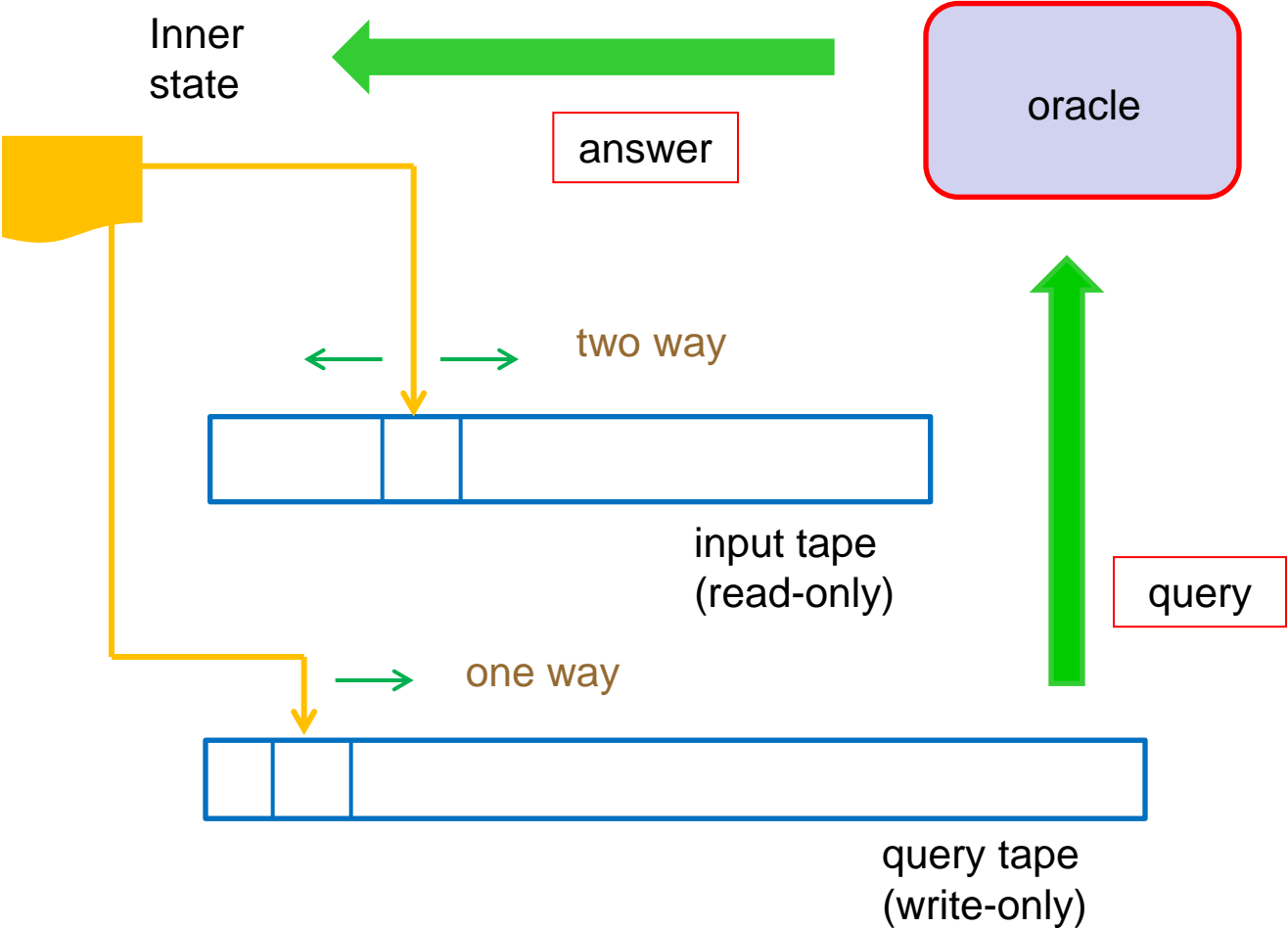
Reductions for Parameterized Problems

- Reductions or reducibility has been so successful to discuss “complete” problems, such as NP-complete problems.
- Now, our goal is to define suitable reductions among parameterized problems in PsubLIN.
- First of all, for a wider range of application, we expand “many-one reduction” to “Turing reduction.”
- To define Turing reduction, we need to introduce a notion of **oracle Turing machine** and a notion of **oracle**.

Oracle Turing Machines I

- Here, we give briefly general notions of oracle Turing machine and oracle.
- (*) The notion of OTMs will be discussed extensively in Week 4.
- An **oracle Turing machine** (OTM) is equipped with an additional tape, called a **query tape**, in which the machine make a query to an oracle.
- An **oracle** is an external information source, which can provide the machine with necessary information via a process of query and answer.

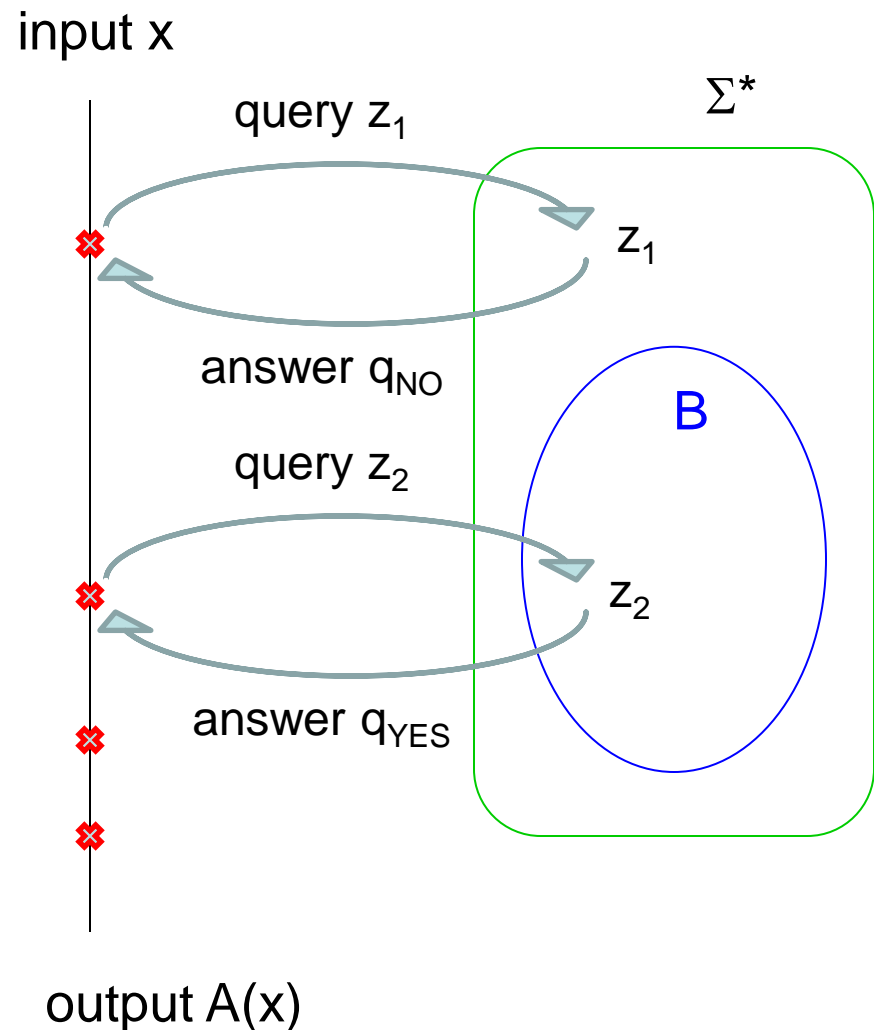
Oracle Turing Machines II



Oracle Computation

- M: OTM for A
- B: oracle

1. M starts with input x .
2. Whenever M writes a query word z on its query tape and enters a query state q_{query} , z is automatically sent to B.
3. The oracle B returns its answer (YES/NO) by changing M's inner state to either q_{yes} or q_{no} .
4. M resumes its computation, starting with q_{yes} or q_{no} .
5. If M halts, output $M(x)$. Otherwise, go to Step 2.



SLRF-T-Reducibility

- We define a notion of (polynomial-time) sub-linear-space reduction family (SLRF).

- $(P_1, m_1) \leq^{\text{SLRF}_T} (P_2, m_2) \Leftrightarrow$

$\forall \varepsilon > 0 \exists M: \text{oracle DTM} \exists l: \text{polylog} \exists k_1, k_2 > 0$ s.t.

1. $M^{P_2}(x)$ runs in $\leq p(|x|)$ time and $\leq m_1(x)^{\varepsilon l(|x|)}$ space
2. Whenever M makes a query to oracle P_2 , M receives its answer and continues a computation.
3. If M make a query z to P_2 , then $m_2(z) \leq m_1(x)^{k_1} + k_1$ and $|z| \leq |x|^{k_2} + k_2$.

- All queried words z have size polynomial in the size of inputs (w.r.t. size parameters).

Short Reductions are Needed

- Unfortunately, in SLRF-T-reduction, query words are too long to make **functional composition** for sub-linear-space machines.
- This raises a serious question whether **PsubLIN may not be closed under \leq_m^L -reductions.**
- This forces us to look for a more restricted notion of reductions to discuss the computational complexity of PsubLIN.
- A simple remedy is to make only “**short**” queries.
- Namely, we demand that **the size of queried word is linear in the size of input** (w.r.t. given size parameters).

Short SLRF-T-Reductions

- We say that (P_1, m_1) is short SLRF-T-reducible to (P_2, m_2) , denoted by $(P_1, m_1) \leq^{\text{sSLRF}_T} (P_2, m_2)$, if the following hold.

This bound is different from SFRF-T-reductions

- $(P_1, m_1) \leq^{\text{sSLRF}_T} (P_2, m_2) \iff$

$\forall \varepsilon > 0 \exists M: \text{oracle } TM \exists l: \text{polylog} \exists k_1, k_2 > 0 \text{ s.t.}$

- $M^{P_2}(x)$ runs in $\leq p(|x|)$ time and $\leq m_1(x)^{\varepsilon l(|x|)}$ space
- Follow the same oracle mechanism
- If M^{P_2} queries z to P_2 , then $m_2(z) \leq k_1 m_1(x) + k_1$ and $|z| \leq |x|^{k_2} + k_2$.

- $A \equiv_r B \iff A \leq_r B \text{ and } B \leq_r A$ for any reduction type r

Comparison of Query Size

oracle machine



input x

query words z



oracle



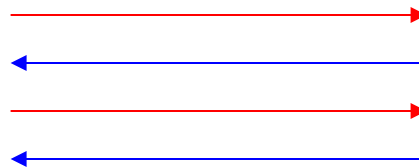
$$m_2(z) \leq m_1(x)^{k_1} + k_1$$

oracle machine



input x

query words z



oracle



$$m_2(z) \leq k_1 m_1(x) + k_1$$

Properties of Short Reductions

- **Proposition:** [Yamakami (2017)]
 - 1) \leq^{SLRF_T} and \leq^{sSLRF_T} : reflexive and transitive.
 - 2) PsubLIN is closed under \leq^{sSLRF_T} -reductions.
 - 3) $\exists X, Y$: recursive s.t. $X \leq^{\text{SLRF}_T} Y$ but $X \not\leq^{\text{sSLRF}_T} Y$.

- **Proposition:** [Yamakami (2017)]

$\forall m \in \{m_{\text{vbl}}, m_{\text{cls}}\} \quad \forall k \geq 3$

- 1) $(2\text{SAT}_{k,m}) \equiv^{\text{sSLRF}_m} (2\text{SAT}_3,m)$
- 2) $(2\text{SAT}_3,m_{\text{vbl}}) \equiv^{\text{sSLRF}_m} (2\text{SAT}_3,m_{\text{cls}})$

However, we don't know if we can replace 2SAT_3 by 2SAT .

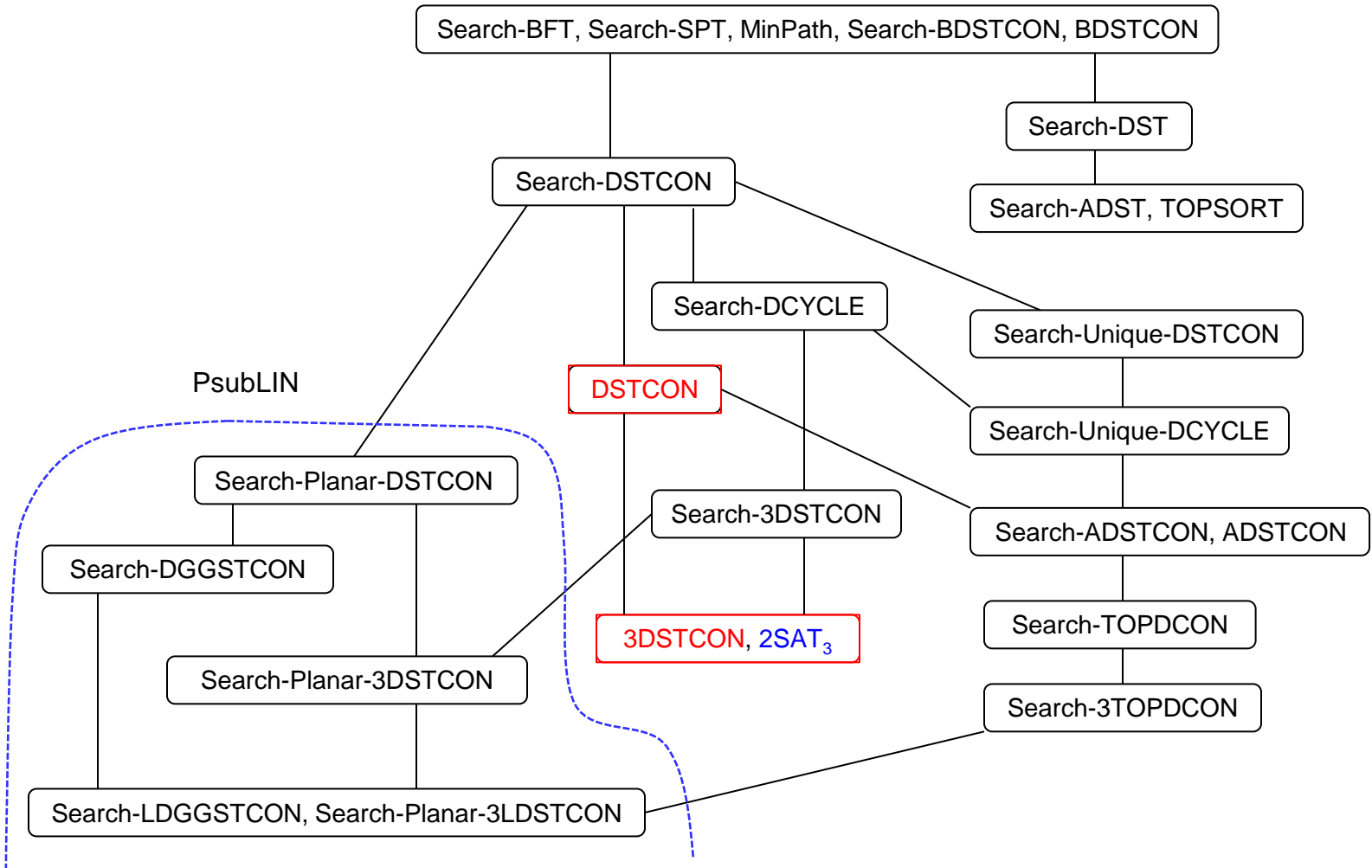
- Hence, it suffices to focus only on $(2\text{SAT}_3,m_{\text{vbl}})$.

Relationships by Short Reductions

- As a simple example of \leq^{sSLRF}_T , let us consider the directed s-t connectivity problem (DSTCON) and its variants.
- The next slide will illustrate certain known relationships among numerous variants of DSTCON problems associated with acyclic graph, planar graph, shortest-path, etc.
- (*) In the next slide, “Search-C” means a search problem in which we are asked to find (and output) a solution to the original decision problem C.

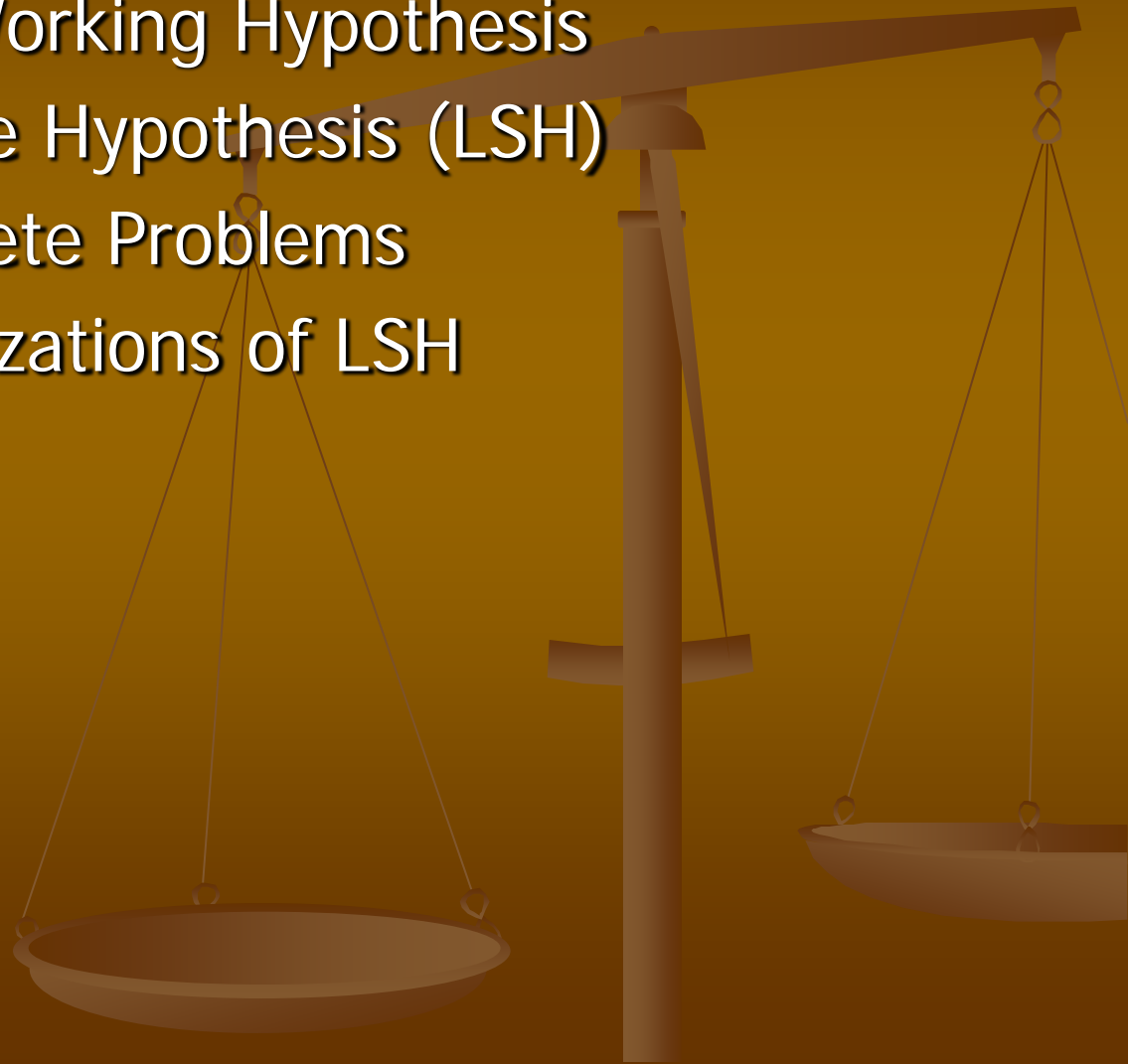
Ordered by sSLRF-reductions

Size parameter: $m_{\text{ver}}(x) = \# \text{ of vertices}$



IV. Linear Space Hypothesis

1. New, Practical Working Hypothesis
2. The Linear Space Hypothesis (LSH)
3. Other NL-Complete Problems
4. Other Characterizations of LSH



New, Practical Working Hypothesis

- As noted earlier, 2SAT with n variables and m clauses is solvable in polynomial time using at most $n^{1-c/\sqrt{\log(n)}} \times \text{polylog}(m+n)$ space.
- However, we do not know whether 2SAT (even 2SAT₃) is solved in polynomial time using $n^\varepsilon \times \text{polylog}(m+n)$ space for a fixed constant $\varepsilon \in [0, 1)$.
- We want to propose a new, practical working hypothesis, which is expected to serve as a driving force to obtain better lower bounds of the computational complexity of various problems.

The Linear Space Hypothesis (LSH) I

- We introduce a working hypothesis called **the linear space hypothesis** (LSH).

- **LSH** (or **LSH for 2SAT₃**) states:

There is no deterministic algorithm that solves 2SAT₃ in time $p(|x|)$ using at most $m_{\text{vbl}}(x)^{\varepsilon l(|x|)}$ space on instance x for a certain polynomial p , a certain polylog function l , and a certain constant $\varepsilon \in [0, 1)$.

- **Open Problem**

Prove or disprove that **LSH for 2SAT₃** \leftrightarrow **LSH for 2SAT**.

The Linear Space Hypothesis (LSH) II

- The previous definition uses the parameterized problem $(2SAT_3, m_{vbl})$. How about $(2SAT_3, m_{cls})$?
- **(Claim)** We can replace m_{vbl} in the above by m_{cls} .

□ Proof Sketch:

This is because $(2SAT_3, m_{vbl}) \equiv^{sSLRF_m} (2SAT_3, m_{cls})$ and PsubLIN is closed under \leq^{sSLRF_m} -reductions.

- **Theorem:** [Yamakmai (2017)]
If LSH for $2SAT_3$ holds, then $L \neq NL$.
- The converse is not yet known.

Other NL-Complete Problems I

- For two column vectors $x = (x_1, x_2, \dots, x_n)^T$ and $y = (y_1, y_2, \dots, y_n)^T$, we define
$$x \geq y \Leftrightarrow x_i \geq y_i \text{ for all index } i \in \{1, 2, \dots, n\}.$$
- $LP_{2,k}$ (linear programming problem)
 - **instance:** a rational $m \times n$ matrix A , a **rational** column vector $b \in \mathbb{Q}^n$, where each row of A has at most two non-zero entries and each column of A has at most k non-zero entries
 - **question:** is there any $\{0, 1\}$ -vector x s.t. $Ax \geq b$?
 - $m_{\text{col}}(x) = \#$ of columns in A
 - $m_{\text{row}}(x) = \#$ of rows in A
- **(Claim)** $LP_{2,k}$ is NL-complete for any $k \geq 3$.

See the next slide.

Other NL-Complete Problems II

$$\begin{array}{cccc} & \mathbf{A} & & \mathbf{x} & & \mathbf{b} \\ \left[\begin{array}{cccc} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{array} \right] & & \left[\begin{array}{c} x_1 \\ x_2 \\ \vdots \\ x_n \end{array} \right] & \geq & \left[\begin{array}{c} b_1 \\ b_2 \\ \vdots \\ b_n \end{array} \right] \end{array}$$

- \mathbf{A} : $m \times n$ matrix
- \mathbf{x} : n -dimensional column vector
- \mathbf{b} : n -dimensional column vector

$$\left\{ \begin{array}{l} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \geq b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \geq b_2 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \geq b_m \end{array} \right.$$

Other Characterizations of LSH

- We have seen $2SAT_3$, $3DSTCON$, and $LP_{2,3}$ so far.
- Interestingly, those three NL-complete problems have a common feature.
- **Theorem:** [Yamakami (2017)]
The following three statements are logically equivalent.
 - LSH for $2SAT_3$ (with m_{vbl} or m_{cls})
 - LSH for $LP_{2,3}$ (with m_{row} or m_{col})
 - LSH for $3DSTCON$ (with m_{ver} or m_{edg})
- **However**, not all NL-complete problems seem to share the above special property concerning LSH.

V. Applications of LSH

1. NL Search Problems
2. Complexity of Search-UOCK
3. NL Optimization Problems
4. Complexity of Max-HPP
5. Topological Sort
6. Complexity of TOPSORT



NL Search Problems

- The first application is in the field of NL search problems.
- **Search-UOCK** (a variant of Knapsack Problem)
 - **instance:** a string w , a sequence (w_1, w_2, \dots, w_n) of strings s.t., $\forall i \in [n]$, if w_i is a substring of w then w_i is unique
 - **solution:** a sequence (i_1, i_2, \dots, i_k) of indices with $k \geq 1$ s.t. $1 \leq i_1 < i_2 < \dots < i_k \leq n$ and $w = w_{i_1} w_{i_2} \dots w_{i_k}$.

	w_1	w_2	w_3	w_4	w_5
input sequence	0010	11	010	0111	1010

input string $w =$ 0 0 1 0 0 1 0 1 0 1 0

||

0010	010	1010
w_1	w_3	w_5

output (1, 3, 5)

Complexity of Search-UOCK

- **Search-UOCK** (again)
 - **instance:** a string w , a sequence (w_1, w_2, \dots, w_n) of strings over alphabet Σ s.t., $\forall i \in [n]$, if w_i is a substring of w then w_i is unique
 - **solution:** a sequence (i_1, i_2, \dots, i_k) of indices with $k \geq 1$ s.t. $1 \leq i_1 < i_2 < \dots < i_k \leq n$ and $w = w_{i_1} w_{i_2} \dots w_{i_k}$.
- **size parameter:** $m_{\text{elm}}(x) = n$ (the number of elements)
- **Theorem:** [Yamakami (2017)]

If LSH (for 2SAT_3) holds, then, for $\forall \varepsilon > 0$, there is no polynomial-time $O(n^{1/2-\varepsilon})$ -space algorithm for (Search-UOCK, m_{elm}).

NL Optimization Problems I

- The second application is in the field of **NL optimization problems**.
- In an optimization problem, intuitively speaking, we are asked to search for optimal solutions satisfying certain predetermined properties for each given input, where “optimality” is measured by **cost functions m** .
- **NLO** = class of NL optimization problems [Tantau (2007), Yamakami (2013)]
- (*) We will discuss optimization problems extensively in Week 9.

NL Optimization Problems II

- We further define an approximation class.
- **LSAS_{NLO}** = class of NLO problems that have log-space approximation schemes [Tantau (2007), Yamakami (2013)]
- A **log-space approximation scheme** for problem P is a DTM M that takes (x,k) as input and outputs a solution y of P using at most $f(k)\log(|x|)$ space with **performance ratio** $R(x,y) \leq 1+1/k$, where $f \in \text{FL}$. Such y is called a **(1+1/k)-approximate solution**.
- **Performance ratio** $R(x,y) = \max\{ |m(x,y)/m^*(x)|, |m^*(x)/m(x,y)| \}$, where $m^*(x) = \max\{ m(x,y) \mid y \text{ is a solution for input } x \}$.

Complexity of Max-HPP

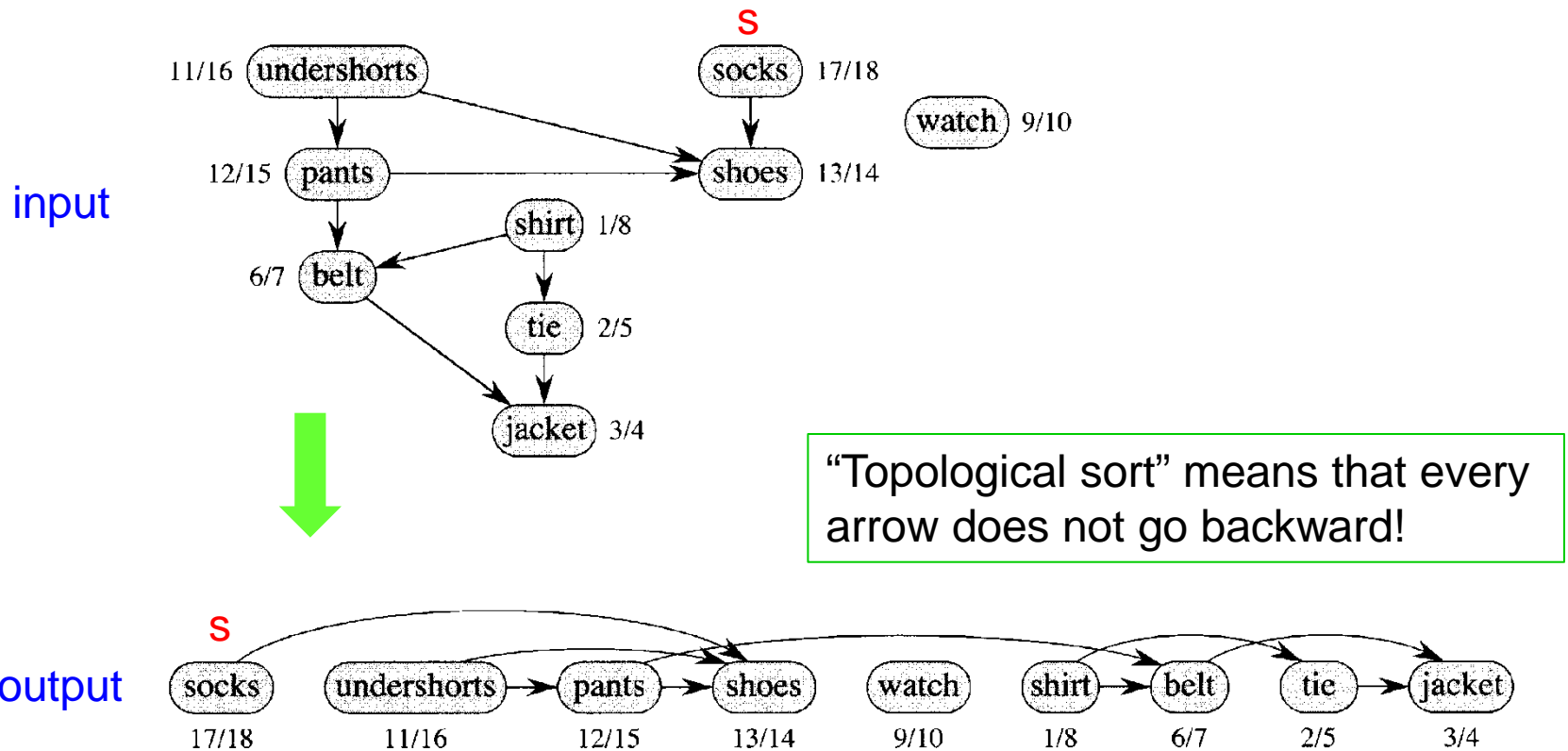
- **Max-HPP (maximum hot potato)** [Tantau (2007)]
 - **instance:** an $n \times n$ matrix A whose entries are in $[n]$, a $d \in [n]$, a start index $i_1 \in [n]$ for $n \in \mathbb{N}^+$
 - **solution:** an index sequence $S = (i_1, i_2, \dots, i_d)$ with $i_j \in [n]$
 - **measure:** total weight
- size parameter: $m_{\text{col}}(x) = n$
- Max-HPP is in LSAS_{NLO} [Tantau (2007)] but it is hard for LO_{NLO} under approximation-preserving exact NC^1 -reduction [Yamakami (2013)].
- **Theorem:** [Yamakami (2017)]

If LSH for 2SAT_3 holds, then, for $\forall \varepsilon > 0$, there is no polynomial-time $O(k^{1/3} \log(m_{\text{col}}(x)))$ -space algorithm finding $(1 + 1/k)$ -approximate solutions of $(\text{Max-HPP}, m_{\text{col}})$.

$$w(S) = \sum_{j=1}^{d-1} A_{i_j, i_{j+1}}$$

Topological Sort

- **Topological sorting problem (TOPSORT)**
 - **instance:** an acyclic directed graph G and a source s in G
 - **output:** a topological sort of G starting from s



Complexity of TOPSORT

- LSH can tell how difficult to solve TOPSORT.
- More precisely, we obtain the following result.
- **Theorem:** [Yamakami (2017)]
If LSH (for $2SAT_3$) holds, then no DTM solves $(TOPSORT, m_{ver})$ in polynomial-time using $O(m_{ver}(x)^{\epsilon/2})$ space on instances x for any fixed constant $\epsilon \in [0, 1)$

Open Problems

- There are numerous problems that have been left unsolved concerning LSH.
- Here are several important questions.
 1. Find more interesting and practical applications of LSH.
 2. Prove or disprove that LSH is true.
 3. Discuss the relationships between LSH for $2SAT_3$ and LSH for $2SAT$.
- (*) We will return to a discussion on LSH in Week 6.



Thank you for listening

Thank you for listening

Q & A

I'm happy to take your question!



END