Theory of Computing Time Complexity

#### Ming-Hsien Tsai

Department of Information Management National Taiwan University

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(original created by Bow-Yaw Wang)

Ming-Hsien Tsai (IM@NTU)

Time Complexity

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- Let us consider  $A = \{0^n 1^n : n \ge 0\}.$
- How much time does a single-tape TM need to decide *A*?
- Consider
  - $M_1 =$  "On input string *w*:
    - Scan the tape and reject if a 0 appears after a 1.
    - Repeat if 0 or 1 appear on the tape:

• Scan across the tape, cross a 0 and a 1.

- If 0's or 1's still remain, reject. Otherwise, accept."
- How much "time" does *M*<sub>1</sub> need for an input *w*?

#### Definition 1

Let *M* be a TM that halts on all inputs. The <u>running time</u> (or <u>time</u> <u>complexity</u>) of *M* is the function  $f : \mathbb{N} \to \mathbb{N}$  where f(n) is the running time of *M* on any input of length *n*.

- If *f*(*n*) is the running time of *M*, we say *M* runs in time *f*(*n*) and *M* is an *f*(*n*) time TM.
- In <u>worst-case analysis</u>, the longest running time of all inputs of a particular length is considered.
- In <u>average-case analysis</u>, the average of all running time of inputs of a particular length is considered instead.
- We only consider worst-case analysis in the course.

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# Big-O and Small-O

#### Definition 2

Let  $f, g : \mathbb{N} \to \mathbb{R}^+$ .  $\underline{f(n) = O(g(n))}$  if there are  $c, n_0 \in \mathbb{Z}^+$  such that for all  $n \ge n_0$ ,

 $f(n) \le c(g(n)).$ 

- g(n) is an <u>upper bound</u> (or an <u>asymptotic upper bound</u>) for f(n).
  n<sup>c</sup>(c ∈ ℝ<sup>+</sup>) is a <u>polynomial bound</u>.
- $2^{n^d}(d \in \mathbb{R}^+)$  is an <u>exponential bound</u>.

#### **Definition 3**

Let  $f, g : \mathbb{N} \to \mathbb{R}^+$ . <u>f (n) = o (g (n))</u> if

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}=0.$$

That is, for any  $c \in \mathbb{R}$ , there is an  $n_0$  that f(n) < c(g) for all  $n \ge n_0$ .

# Time Complexity of *M*<sub>1</sub>

- Recall  $M_1 = "On input string w:$ 
  - Scan the tape and reject if a 0 appears after a 1.
  - Repeat if 0 or 1 appear on the tape:

Scan across the tape, cross a 0 and a 1.

- If 0's or 1's still remain, reject. Otherwise, accept."
- Let |w| = n.
  - Step 1 takes O(n) (precisely,  $\leq n$ ).
  - Step 2 has O(n) iterations (precisely,  $\leq n/2$ ).
    - ★ An iteration takes O(n) (precisely,  $\leq n$ ).
  - Step 3 takes O(n) (precisely,  $\leq n$ ).
- The TM  $M_1$  decides  $A = \{0^n 1^n : n \ge 0\}$  in time  $O(n^2)$ .
  - $\blacktriangleright O(n^2) = O(n) + O(n) \times O(n) + O(n).$

#### Definition 4

Let  $t : \mathbb{N} \to \mathbb{R}^+$ . The time complexity class TIME(t(n)) is the collection of all languages that are decided by a O(t(n)) time TM.

•  $A = \{0^n 1^n : n \ge 0\}$  is decided by  $M_1$  in time  $O(n^2)$ .  $A \in TIME(n^2)$ .

- Time complexity classes characterizes languages, not TM's.
  - We don't say  $M_1 \in TIME(n^2)$ .
- A language may be decided by several TM's.
- Can *A* be decided more quickly asymptotically?

# Models and Time Complexity

- Consider the following TM:  $M_2 =$  "On input string *w*:
  - Scan the tape and reject if a 0 appears after a 1.
  - Repeat if 0 or 1 appear on the tape:
    - Scan the tape and check if the total number of 0's and 1's is even. If not, reject.
    - Scan the tape, cross every other 0 from the first 0, and cross every other 1 from the first 1.
  - If 0's or 1's still remain, reject. Otherwise, accept."
- Analysis of M<sub>2</sub>.
  - ▶ Step 1 takes *O*(*n*).
  - Step 2 has  $O(\lg n) (= \log_2(n))$  iterations (why?). At each iteration,
    - ★ Step 1 takes O(n).
    - **\*** Step 2 takes O(n).
  - ► Step 3 takes *O*(*n*).
- $M_2$  decides A in time  $O(n \lg n)$ .
  - $O(n \lg n) = O(n) + O(\lg n) \times O(n) + O(n).$

# Models and Time Complexity

- Consider the following two-tape TM:
  - $M_3 =$  "On input string *w*:
    - **1** Scan tape 1 and reject if a 0 appears after a 1.
    - Scan tape 1 and copy the 0's onto tape 2.
    - Scan tape 1 and cross a 0 on tape 2 for a 1 on tape 1.
    - If all 0's are crossed off before reading all 1's, reject. If some 0's are left after reading all 1's, reject. Otherwise, accept."
- Analysis of *M*<sub>3</sub>.
  - ► Each step takes *O*(*n*).
- For the same language  $A = \{0^n 1^n : n \ge 0\}$ .
  - The TM  $M_1$  decides A in time  $O(n^2)$ , the TM  $M_2$  decides A in time  $O(n \lg n)$ , and the two-tape  $M_3$  decides A in time O(n).
- In computability theory, all reasonable variants of TM's decide the same language (Church-Turing thesis).
- In complexity theory, different variants of TM's may decide the same in different time.

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# Complexity Relationship with Multitape TM's

#### Theorem 5

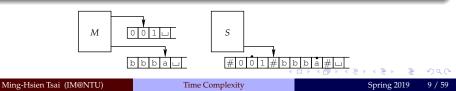
*Let* t(n) *be a function with*  $t(n) \ge n$ *. Every* t(n) *time multitape Turing machine has an equivalent*  $O(t^2(n))$  *time single-tape TM.* 

#### Proof.

We analyze the simulation of a *k*-tape TM *M* is by the TM *S*. Observe that each tape of *M* has length at most t(n) (why?). For each step of *M*, *S* has two passes:

- The first pass gathers information (*O*(*kt*(*n*))).
- The second pass updates information with at most k shifts  $(O(k^2t(n)))$ .

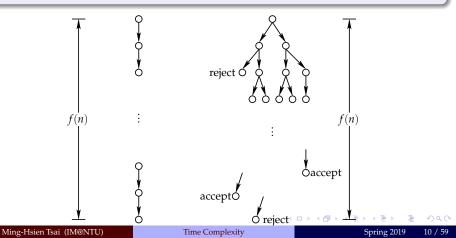
Hence *S* takes  $O(n) + O(k^2t^2(n)) (= O(n) + O(t(n)) \times O(k^2t(n)))$ . Since  $t(n) \ge n$ , we have *S* runs in time  $O(t^2(n))$  (*k* is independent of the input).



# Time Complexity of Nondterministic TM's

#### **Definition 6**

Let *N* be a nondeterministic TM that is a decider. The <u>running time</u> of *N* is a function  $f : \mathbb{N} \to \mathbb{N}$  where f(n) is the maximum number of steps among any branch of *N*'s computation on input of length *n*.



# Complexity Relationship with NTM's

#### Theorem 7

Let t(n) be a function with  $t(n) \ge n$ . Every t(n) time single-tape NTM has an equivalent  $2^{O(t(n))}$  time single-tape TM.

#### Proof.

Let *N* be an NTM running in time t(n). Recall the simulation of *N* by a 3-tape TM *D* with the address tape alphabet  $\Sigma_b = \{1, 2, ..., b\}$  (*b* is the maximal number of choices allowed in *N*).

Since *N* runs in time t(n), the computation tree of *N* has  $O(b^{t(n)})$  nodes. For each node, *D* simulates it from the start configuration and thus takes time O(t(n)). Hence the simulation of *N* on the 3-tape *D* takes  $2^{O(t(n))}(=O(t(n)) \times O(b^{t(n)}))$  time.

By Theorem 5, *D* can be simulated by a single-tape TM in time  $(2^{O(t(n))})^2 = 2^{O(t(n))}$ .

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# The Class P

- It turns out that reasonable deterministic variants of TM's can be simulated by a TM with a polynomial time overhead.
  - ▶ multitape TM's, TM's with random access memory, etc.
- The polynomial time complexity class is rather robust.
  - That is, it remains the same with different computational models.

#### **Definition 8**

*P* is the class of languages decidable in polynomial time on a determinsitic single-tape TM. That is,

$$P = \bigcup_k TIME(n^k).$$

- We are interested in intrinsic characters of computation and hence ignore the difference among variants of TM's in this course.
- Solving a problem in time O(n) and  $O(n^{100})$  certainly makes lots of difference in practice.

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- We will give some problems and their deciders in *P*.
- We will use a reasonable encoding for our problems.
- For natural numbers, we represent them in base  $k \ge 2$  notation.
  - Uniary encoding is not reasonable.
  - $17 = 10001_2 = 11111111111111111_1$ .
- For graphs, we use <u>adjacency matrices</u>.
  - ► The (*i*, *j*)th entry is 1 if there is an edge from node *i* to node *j*.
- Note that the size of a graph is a polynomial in our reasonable encoding of the graph.
- Consider the following problem:

 $PATH = \{ \langle G, s, t \rangle : \text{ there is a path from } s \text{ to } t \text{ in the directed graph } G \}.$ 

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Theorem 9	
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 $PATH \in P.$ 

### Proof.

#### Consider

M = "On input  $\langle G, s, t \rangle$  where *G* is a directed graph with nodes *s* and *t*:

- Place a mark on s.
- Repeat until no more nodes are marked:
  - If there is an edge (a, b) from a marked node a to an unmarked node b, mark b.
- If *t* is marked, accept. Otherwise, reject."

Let *m* be the number of nodes in *G*. Step 1 takes O(1). Step 2 has O(m) iterations; each iteration takes  $O(m^2)$ . Step 3 take O(1). Hence *M* is a polynomial time algorithm for *PATH*.

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# RELPRIME is in P

• Consider *RELPRIME* = { $\langle x, y \rangle$  : *x* and *y* are relatively prime}.

### Theorem 10

 $RELPRIME \in P.$ 

### Proof.

Recall the Euclidean algorithm:

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E = "On input \langle x, y \rangle where x, y \in \mathbb{Z}^+ in binary:
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• Repeat until y = 0
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  x \leftarrow x \mod y.
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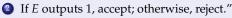
2 Exchange *x* and *y*.

```
Output x".
```

Consider

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R = "On input \langle x, y \rangle where x, y \in \mathbb{Z}^+ in binary:
```

**1** Run *E* on  $\langle x, y \rangle$ .



#### Proof (cont'd).

It remains to analyze the time complexity of *E*. We show each execution (except perhaps the first) of Step 1.1 cuts the value of *x* by a half.

Observe that x > y for each execution (except perhaps the first). Right before Step 1.1,

- If  $\frac{x}{2} \ge y$ , then  $x \mod y < y \le \frac{x}{2}$ .
- If  $\frac{x}{2} < y$ , then  $x \mod y = x y \le \frac{x}{2}$ .

Hence *x* draps by a half right after Step 1.1. Since *x* and *y* exchanges values at Step 1.2, *x* and *y* drop by a half every other iteration. Hence Step 1 has  $O(\lg x + \lg y)$  iterations. Recall we are use the binary encoding. Step 1 has a polynomial number of iterations. Moreover, comparison, modulo, and assignments take a polynomial number of time. *R* decides *RELPRIME* in polynomial time.

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# Context-Free Languages are in P

#### Theorem 11

Every context-free language is a member of P.

### Proof.

Consider D = "On input  $w = w_1 w_2 \cdots w_n$ :

1 If 
$$w = \epsilon$$
 and  $S \longrightarrow \epsilon \in G$ , accept.

For each variable A

2 For 
$$i = 1$$
 to  $n$  do

// *G* is in Chomsky normal form

// Handle substrings of length 1

If  $A \longrightarrow b \in G$  and  $w_i = b$ , put A in table(i, i).

So For l = 2 to n do

// Handle substrings of length *l* 

• For i = 1 to n - l + 1 do // Handle the substring at  $w_i w_{i+1} \cdots w_j$ 

• 
$$j = i + l - 1$$
.  
• For  $k = i$  to  $j - 1$  do  
For each  $A \longrightarrow BC \in G$  do  
If  $B \in table(i, k)$  and  $C \in table(k + 1, j)$ , add  $A$  to  $table(i, j)$ .

If  $S \in table(1, n)$ , accept; otherwise, reject."

#### Proof.

Let *G* have *v* variables and *r* rules (v, r are independent of |w|).

- Step 1 executes once.
- Step 2 has *O*(*n*) iterations.
  - Each iteration has O(v) subiterations.
- Step 3 has *O*(*n*) iterations.
  - Each iteration has O(n) subiterations.
    - Each subiteration has O(n) subsubiterations; each subsubiterations has r subsubsubiterations.
- Step 4 executes once.

Since table lookup and insertion take a polynomial number of time, *D* runs in polynomial time.

# The Class NP

#### Definition 12

A verifier for a language *A* is an algorithm *V* where

 $A = \{w : V \text{ accepts } \langle w, c \rangle \text{ for some } c\}.$ 

*c* is a certificate or proof of membership in *A*. A polynomial time verifier runs in polynomial time in the length of w (not  $\langle w, c \rangle$ ). A language *A* is polynomially verifiable if it has a polynomial time verifier.

- Note that a certificate has a length polynomial in |w|.
  - Otherwise, *V* cannot run in polynomial time in |w|.



- A <u>Hamiltonian path</u> in a directed graph *G* is a path that goes through every node exactly once. Consider
  - HAMPATH ={ $\langle G, s, t \rangle$  : *G* is a directed graph with a Hamiltonian path from *s* to *t*}.
- $HAMPATH \in NP$ .
  - Verifying whether c is a Hamiltonian path from s to t can be done in polynomial time.
  - A certificate for  $\langle G, s, t \rangle \in HAMPATH$  is a Hamiltonian path from *s* to *t*.
- Finding a Hamiltonian path from *s* to *t* seems harder.

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• A natural number is <u>composite</u> if it is the product of two integers greater than 1. Consider

 $COMPOSITES = \{ \langle x \rangle : x = pq \text{ for } p, q \in \mathbb{Z}^+ \text{ and } p, q > 1 \}.$ 

- $COMPOSITES \in NP$ .
  - ► Verifying whether x is the product of two integers p, q with p, q > 1 can be done in polynomial time.
  - A certificate for  $\langle x \rangle \in COMPOSITES$  is  $\langle p, q \rangle$  with  $p, q \in \mathbb{Z}^+$ , p, q > 1, and x = pq.
- Finding  $p, q \in \mathbb{Z}^+$  with  $p, q \in \mathbb{Z}^+$ , p, q > 1, and x = pq seems harder.
  - Can we find *p* in  $\{2, 3, ..., \lceil \sqrt{x} \rceil\}$ ? Is it a polynomial time algorithm?
- There is however a polynomial time algorithm for *COMPOSITES*.

- Consider
  - $N_1 =$
  - "On input  $\langle G, s, t \rangle$  where *G* is a directed graph with nodes *s*, *t*:
    - **(**) Nondeterministically write *m* numbers  $p_1, \ldots, p_m$  (*G* has *m* nodes).
    - If there is any repetition, reject.
    - 3 If  $s \neq p_1$  or  $t \neq p_m$ , reject.
    - If  $(p_i, p_{i+1})$  is not an edge of *G* for some  $1 \le i < m$ , reject.
    - Otherwise, accept."
- Since  $N_1$  runs in polynomial time and  $L(N_1) = HAMPATH$ ,  $N_1$  decides HAMPATH.

# NP and NTM's

#### Theorem 14

*A language is in NP if and only if it is decided by a nondeterministic polynomial time Turing machine.* 

#### Proof.

Let *V* be a verifier for a language *A* running in time  $n^k$ . Consider N = "On input *w* of length *n*:

- Nondeterministically select string *c* of length  $\leq n^k$ .
- 2 Run *V* on  $\langle w, c \rangle$ .
- 3 If *V* accepts, accept; otherwise, reject."

Conversely, let the NTM *N* decide *A* and *c* the address of an accepting configuration in the computation tree of *N*. Consider

- V = "On input  $\langle w, c \rangle$ :
  - 1 9
    - Simulate N on w from the start configuration by c.

If the configuration with address *c* is accepting, accept; otherwise, reject."

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### The Nondeterministic Time Complexity Class

# Definition 15 NTIME (t (n)) = { L : L is a language decided by a O (t (n)) time NTM }.

Corollary 16

$$NP = \bigcup_k NTIME(n^k).$$

• Recall that class *TIME*(*t*(*n*)) and

$$P = \bigcup_k TIME(n^k).$$

# Cliques in Graphs

- A <u>clique</u> in an undirected graph is subgraph where every two nodes are connected.
- A *k*-clique is a clique with *k* nodes.
- Consider

 $CLIQUE = \{ \langle G, k \rangle : G \text{ is an undirected graph with a } k\text{-clique} \}.$ 

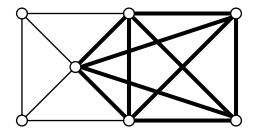


Figure: A Graph has a 5-Clique

# CLIQUE is in NP

Theorem 17
$CLIQUE \in NP.$

Proof.

Consider

- V = "On input  $\langle \langle G, k \rangle, c \rangle$ :
  - Check if *c* is a set of *k* nodes in *G*.
  - Output: Check if G contains all edges connecting nodes in c.
  - If both pass, accept; otherwise, reject."

Alternatively, consider

- N = "On input  $\langle G, k \rangle$  where *G* is an undirected graph:
  - Nondeterministically select a set *c* of *k* nodes of *G*.
  - 2 Check if *G* contains all edges connecting nodes in *c*.
  - If yes, accept; otherwise, reject."

## Subset Sums

#### Consider

$$SUBSETSUM = \{ \langle S, t \rangle : S = \{x_1, \dots, x_k\} \text{ a multiset and for some} \\ \{y_1, \dots, y_l\} \subseteq S, \Sigma y_i = t \}$$

### Theorem 18

 $SUBSETSUM \in NP.$ 

### Proof.

- $V = "On input \langle \langle S, t \rangle, c \rangle:$ 
  - Check if S contains all the numbers in c.
  - Output: Check if c sums to t.
  - If both pass, accept; otherwise, reject."

Alternatively, N = "On input  $\langle S, t \rangle$  where *S* is a multiset:

- Nondeterministcally select a subset *c* of numbers from *S*.
- 2 Check if c sums to t.
- If both pass, accept; otherwise, reject."

Definition 19  $coNP = \{L : \overline{L} \in NP\}.$ 

- $\overline{HAMPATH} \in coNP$  since  $\overline{HAMPATH} = HAMPATH \in NP$ .
  - ► *HAMPATH* does not appear to be polynomial time verifiable.
  - What is a certificate showing there is no Hamiltonian path?
- $\overline{CLIQUE} \in coNP$ .
  - *CLIQUE* does not appear to be polynomial time verifiable.
  - What is a certificate showing there is no k-clique?
- $\overline{SUBSETSUM} \in coNP$ .
  - ▶ *SUBSETSUM* does not appear to be polynomial time verifiable.
  - What is a certificate showing there is no subset summing to *t*?
- We do not know if *coNP* is different from *NP*.

- Recall
  - ► *P* is the class of languages which membership can be decided quickly.
  - ▶ *NP* is the class of languages which membership can be verified quickly.
- We have shown *HAMPATH*, *CLIQUE*, *SUBSETSUM*  $\in$  *NP*.
- We have shown PATH,  $RELPRIME \in P$ .
- Since a polynomial time TM is also a polynomial NTM, we have

 $L \in P$  implies  $L \in NP$  for every language L.

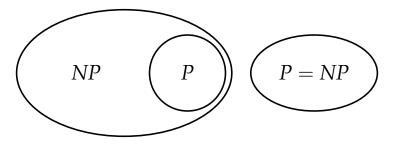


Figure: Possible Relation between P and NP

• To the best of our knowledge, we only know

$$NP \subseteq EXPTIME = \bigcup_{k} TIME(2^{n^{k}}).$$
 (Theorem 7)

• Particularly, we do no know if  $P \stackrel{?}{=} NP$ .

## Satisfiability

- Let  $\mathbb{B} = \{0, 1\}$  be the <u>truth values</u>.
- A Boolean variable takes values from B.
- Recall the Boolean operations

• A <u>Boolean formula</u> is an expression constructed from Boolean variables and opearations.

•  $\phi = (\overline{x} \land y) \lor (x \land \overline{z})$  is a Boolean formula.

- A Boolean formula is <u>satisfiable</u> if an assignments of 0's and 1's to Boolean variables makes the formula evaluate to 1.
  - $\phi$  is satisfiable by taking  $\{x \mapsto \mathbf{0}, y \mapsto \mathbf{1}, z \mapsto \mathbf{0}\}$ .

### The Satisfiability Problem

- The <u>satisfiability problem</u> is to test whether a Boolean formula is satisfiable.
- Consider

 $SAT = \{ \langle \phi \rangle : \phi \text{ is a satisfiable Boolean formula} \}.$ 

Theorem 20 (Cook-Levin)  $SAT \in P$  if and only if P = NP.





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# Polynomial Time Reducibility

### Definition 21

 $f: \Sigma^* \to \Sigma^*$  is a <u>polynomial time computable function</u> if a polynomial time TM *M* halts with only f(w) on its tape upon any input *w*.

### Definition 22

A language *A* is polynomial time mapping reducible (polynomial time reducible, or polynomial time many-one reducible) to a language *B* (written  $A \leq_P B$ ) if there is a polynomial time computable function  $f: \Sigma^* \to \Sigma^*$  that

```
w \in A if and only if f(w) \in B for every w.
```

*f* is called the polynomial time reduction of *A* to *B*.

• Recall the definitions of computable functions and mapping reducibility.

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# Properties about Polynomial Time Reducibility

Theorem 23

If  $A \leq_P B$  and  $B \in P$ ,  $A \in P$ .

#### Proof.

Let the TM *M* decide *B* and *f* a polynomial time reduction of *A* to *B*. Consider

N = "On input w:

- Compute f(w).
- 2 Run M on f(w)."

Since the composition of two polynomials is again a polynomial, *N* runs in polynomial time.

# The 3SAT Problem

- A <u>literal</u> is a Boolean variable or its negation.
- A clause is a disjunction ( $\lor$ ) of literals.
  - $x_1 \lor \overline{x_2} \lor \overline{x_3} \lor x_4$  is a clause.
- A Boolean formula is in <u>conjunctive normal form</u> (or a CNF-formula) if it is a conjunction (∧) of clauses.
  - $(x_1 \lor \overline{x_2} \lor \overline{x_3} \lor x_4) \land (x_2 \lor x_2 \lor \overline{x_5}) \land (x_4 \lor x_6)$  is a CNF-formula.
- In a satisfiable CNF-formula, each clause must contain at least one literal assigned to 1.
- A Boolean formula is a <u>3CNF-formula</u> if it is a CNF-formula whose clauses have three literals.
  - $(x_1 \lor \overline{x_3} \lor x_4) \land (x_2 \lor x_2 \lor \overline{x_5}) \land (x_4 \lor x_5 \lor \overline{x_6})$  is a 3CNF-formula.

Consider

 $3SAT = \{ \langle \phi \rangle : \phi \text{ is a satisfiable 3CNF-formula} \}.$ 

# $3SAT \leq_P CLIQUE$

#### Theorem 24

 $3SAT \leq_P CLIQUE.$ 

#### Proof.

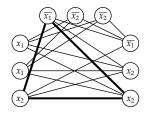
Given a 3CNF-formula  $\phi = (a_1 \lor b_1 \lor c_1) \land (a_2 \lor b_2 \lor c_2) \land \cdots \land (a_k \lor b_k \lor c_k)$ , we would like to find a graph *G* and a number *k* such that  $\langle \phi \rangle \in 3SAT$  if and only if  $\langle G, k \rangle \in CLIQUE$ . We need gadgets to simulate Boolean variables and clauses in  $\phi$ .

- For each clause  $a_i \lor b_i \lor c_i$ , add three corresponding nodes to *G*.
  - G has 3k nodes.
- For each pair of nodes in *G*, add an edge except when
  - the pair of nodes correspond to literals in a clause.
  - the pair of nodes correspond to complementary literals.

We next show that  $\phi$  is satisfiable if and only if *G* has a *k*-clique.

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# $3SAT \leq_P CLIQUE$



 $(x_1 \lor x_1 \lor x_2) \land (\overline{x_1} \lor \overline{x_2} \lor \overline{x_2}) \land (\overline{x_1} \lor x_2 \lor x_2)$ 

### Proof.

Suppose  $\phi$  has a satisfying assignment. Each clause has at least one literal assigned to 1. We pick a node corresponding to true literal from each clause. Any pair of the chosen nodes do not belong to the same clause. Since a literal and its complement cannot be 1 simultaneously, any pair of the chosen nodes are not complementary. Hence there is an edge between any pair of the chosen nodes. We have a *k*-clique. Conversely, suppose there is a *k*-clique. Since there is no edge between any two nodes in a clause, the *k*-clique must have one node from each of the *k* clauses. Moreover, there is no edge between complementary literals. Either a literal or its complement appears in the *k*-clique but not both.  $\phi$  is satisfied by the assignment making literals in the clique true.

It is easy to see that *G* can be constructed from  $\phi$  in polynomial time.

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# **NP-Completeness**

### Definition 25

### A language *B* is <u>NP-complete</u> if

- *B* is in *NP*; and
- every *A* in *NP* is polynomial time reducible to *B*.

#### Theorem 26

*If B is NP-complete and*  $B \in P$ *, then* P = NP*.* 

### Theorem 27

*If*  $C \in NP$ , *B is NP-complete*, *and*  $B \leq_P C$ , *then C is NP-complete*.

#### Proof.

Since *B* is *NP*-complete, there is a polynomial time reduction *f* of *A* to *B* for any  $A \in NP$ . Since  $B \leq_P C$ , there is a polynomial time reduction *g* of *B* to *C*.  $g \circ f$  is a polynomial time reduction of *A* to *C*.

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#### Theorem 28

SAT is NP-complete.

### Proof.

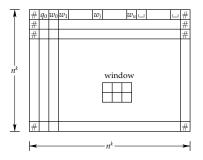
For any Boolean formula  $\phi$ , an NTM nondeterministically choose a truth assignment. It checks whether the assignment satisfies  $\phi$ . If so, accept; otherwise, reject. Hence  $SAT \in NP$ .

Let  $A \in NP$  and the NTM N decide A in  $n^k$  time. For any input w, a <u>tableau</u> for N on w is an  $n^k \times n^k$  table whose rows are the configurations along a branch of the computation of N on w. A tableau of size  $n^k \times n^k$  has  $n^k \times n^k$  cells. We assume each configuration starts and ends with a # symbol. A tableau is <u>accepting</u> if any of its rows is an accepting configuration.

Each accepting tableau for *N* on *w* corresponds to an accepting computation of *N* on *w*. We therefore construct a Boolean formula  $\phi$  such that  $\phi$  is satisfiable if and only if there is an accepting tableau for *N* on *w*.

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## **Cook-Levin Theorem**



#### Proof (cont'd).

Let  $C = Q \cup \Gamma \cup \{\#\}$  where Q and  $\Gamma$  are the states and the tape alphabet of N. For  $1 \le i, j \le n^k$  and  $s \in C$ , the Boolean variable  $x_{i,j,s}$  denotes the content of the cell *cell*[i, j]. That is,  $x_{i,j,s}$  is 1 if and only if *cell*[i, j] = s. To force each cell to contain exactly one symbol from C, consider

$$\phi_{\text{cell}} = \bigwedge_{1 \le i, j \le n^k} \left[ \left( \bigvee_{s \in C} x_{i, j, s} \right) \land \left( \bigwedge_{s, t \in C, s \neq t} (\overline{x_{i, j, s}} \lor \overline{x_{i, j, t}}) \right) \right]$$

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### Proof (cont'd).

To force the tableau to begin with the start configuration, consider

$$\phi_{\text{start}} = x_{1,1,\#} \wedge x_{1,2,q_0} \wedge \\ x_{1,3,w_1} \wedge x_{1,4,w_2} \wedge \dots \wedge x_{1,n+2,w_n} \wedge \\ x_{1,n+3,\square} \wedge \dots \wedge x_{1,n^{k}-1,\square} \wedge x_{1,n^{k},\#}.$$

To force an accepting configuration to appear in the tableau, consider

$$\phi_{\mathrm{accept}} = \bigvee_{1 \leq i,j \leq n^k} x_{i,j,q_{\mathrm{accept}}}.$$

To force the configuration at row *i* yields the configuration at row *i* + 1, consider a window of 2 × 3 cells. For example, assume  $\delta(q_1, a) = \{(q_1, b, R)\}$  and  $\overline{\delta(q_1, b)} = \{(q_2, c, L), (q_2, a, R)\}$ . The following windows are valid:

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#### Proof.

Since *C* is finite, there are only a finite number of valid windows. For any window *W*  $\frac{c_1}{c_4}$   $\frac{c_2}{c_5}$   $\frac{c_3}{c_6}$ , consider

 $\psi_{W} = x_{i,j-1,c_{1}} \land x_{i,j,c_{2}} \land x_{i,j+1,c_{3}} \land x_{i+1,j-1,c_{4}} \land x_{i+1,j,c_{5}} \land x_{i+1,j+1,c_{6}}$ 

To force every window in the tableau to be valid, consider

$$\phi_{\text{move}} = \bigwedge_{1 \le i \le n^k, 1 \le j < n^k} \left( \bigvee_{\text{W is a valid}} \psi_{\text{W}} \right).$$

Finally, consider the following Boolean formula:

 $\phi = \phi_{\text{cell}} \land \phi_{\text{start}} \land \phi_{\text{accept}} \land \phi_{\text{move}}.$ 

 $|\phi_{\text{cell}}| = O(n^{2k}), |\phi_{\text{start}}| = O(n^k), |\phi_{\text{accept}}| = O(n^{2k}), \text{ and } |\phi_{\text{move}}| = O(n^{2k}).$  Hence  $|\phi| = O(n^{2k})$ . Moreover,  $\phi$  can be constructed from N in time polynomial in n.

# 3SAT is NP-Complete

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### Corollary 29

3SAT is NP-complete.

### Proof.

We convert the Boolean formula  $\phi$  in the proof of Theorem 28 into a 3CNF-formula. We begin by converting  $\phi$  into a CNF-formula.

Observe that the conjunction of CNF-formulae is again a CNF-formula. Note that  $\phi_{cell}$ ,  $\phi_{start}$ , and  $\phi_{accept}$  are already in CNF (why?).  $\phi_{move}$  is of the following form:

$$\bigwedge_{\leq i \leq n^k, 1 \leq j < n^k} \left( \bigvee_{\text{W is valid}} (l_1 \land l_2 \land l_3 \land l_4 \land l_5 \land l_6) \right)$$

By the law of distribution,  $\phi_{\text{move}}$  can be converted into a CNF-formula. Note that the conversion may increase the size of  $\phi_{\text{move}}$ . Yet the size is independent of |w|. Hence the size of the CNF-formula  $\phi$  still polynomial in |w|. To a clause of *k* literals into clauses of 3 literals, consider  $l_1 \mapsto (l_1 \vee l_1 \vee l_1)$ ,  $l_1 \vee l_2 \mapsto (l_1 \vee l_2 \vee l_2)$ , and  $l_1 \vee l_2 \vee \cdots \wedge l_p \mapsto (l_1 \vee l_2 \vee z_1) \wedge (\overline{z_1} \vee l_3 \vee z_2) \wedge \cdots \wedge (\overline{z_{p-3}} \vee l_{p-1} \vee l_p)$ .

- To find more *NP*-complete problems, we apply Theorem 27.
- Concretely, to show C is NP-complete, do
  - prove C is in NP; and
  - ▶ find a polynomial time reduction of an *NP*-complete problem (say, *3SAT*) to *C*.
- In Theorem 24, we have shown  $3SAT \leq_P CLIQUE$ . Therefore

Corollary 30 CLIQUE is NP-complete.

- Let  $\phi$  be a 3CNF-formula.
  - For instance  $(x_0 \lor \overline{x}_1 \lor x_2) \land (\overline{x}_0 \lor x_1 \lor \overline{x}_3)$ .
- An <u>≠-assignment</u> to the variables of φ makes every clause of φ to have two literals with different truth values.
  - ► { $x_0 \mapsto 1, x_1 \mapsto 0, x_2 \mapsto 0, x_3 \mapsto 0$ } is an  $\neq$ -assignment.
  - ► { $x_0 \mapsto 1, x_1 \mapsto 0, x_2 \mapsto 1, x_3 \mapsto 0$ } is not an  $\neq$ -assignment.
- Observe that the negation of an ≠-assignment is again an ≠-assignment.
  - $\{x_0 \mapsto \mathbf{0}, x_1 \mapsto \mathbf{1}, x_2 \mapsto \mathbf{1}, x_3 \mapsto \mathbf{1}\}$  is an  $\neq$ -assignment.

Define

 $\neq$  *SAT* = { $\langle \phi \rangle$  :  $\phi$  is a 3CNF formula with an  $\neq$ -assignment}.

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# $\neq$ *SAT* is *NP*-Complete

#### Theorem 31

 $\neq$  SAT is NP-complete.

### Proof.

We reduce 3SAT to  $\neq SAT$ . Let *b* be a Boolean variable not in  $\phi$ . Replace the *i*-th clause  $(l_{i0} \lor l_{i1} \lor l_{i2})$  in  $\phi$  with  $(l_{i0} \lor l_{i1} \lor z_i)$  and  $(\overline{z}_i \lor l_{i2} \lor b)$ .

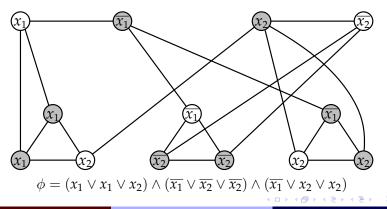
Assume  $\phi$  is satisfiable. We extend the satisfying assignment by assigning *b* to 0 and  $z_i$  to  $\begin{cases} 0 & \text{if } l_{i2} = 0 \text{ or } l_{i0} = l_{i1} = l_{i2} = 1 \\ 1 & \text{otherwise} \end{cases}$ .

Conversely, assume we have an  $\neq$ -assignment. We show the  $\neq$ -assignment satisfies  $\phi$  as well. Without loss of generality, assume b = 0 in the  $\neq$ -assignment. If  $l_{i2} = 1$  in the  $\neq$ -assignment, the *i*-th clause in  $\phi$  is satisfied. Otherwise,  $\overline{z}_i = 1$  for  $\overline{z}_i \lor l_{i2} \lor b$  has two literals of different values. Then  $l_{i0} = 1$  or  $l_{i1} = 1$  because  $l_{i0} \lor l_{i1} \lor z_i$  has two literals of different values. The *i*-th clause of  $\phi$  is satisfied as well.

## Vertex Cover

- Let *G* be an undirected graph. A <u>vertex cover</u> of *G* is a set of nodes where every edge of *G* touches one of these nodes.
- Consider

 $VERTEXCOVER = \{ \langle G, k \rangle : G \text{ is an undirected graph with} a k-node vertex cover \}.$ 



# VERTEXCOVER is NP-Complete

### Theorem 32 VERTEXCOVER is NP-complete.

### Proof.

We show  $3SAT \leq_P VERTEXCOVER$ . Let  $\phi$  be a 3CNF-formula with l variables and k clauses. We need gadgets to simulate variables and clauses in  $\phi$ . We want to find a graph G such that  $\phi$  is satisfiable if and only if G has a l + 2k-node vertex cover.

- For each variable *x*, add two nodes corresponding to *x* and  $\overline{x}$ . Moreover, an edge connecting the two nodes is added to *G*.
- For each clause, add three nodes corresponding to the literals. Three edges connecting the three nodes are added to *G*. Moreover, each of the three node is connected to the node of identical label in the variable gadgets.

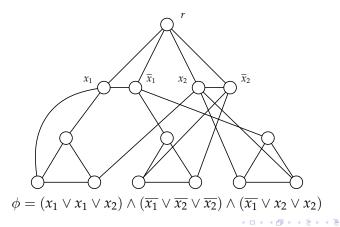
### Proof (cont'd).

We show that  $\phi$  is satisfiable if and only if *G* has an (l + 2k)-node vertex cover. Suppose there is a truth assignment satisfying  $\phi$ . Either of the nodes in each variable gadget is 1. We select these *l* nodes. For each clause gadget, one of the three nodes corresponds to a true literal. We select the other two nodes. These l + 2k nodes form a vertex cover. Conversely, suppose *G* has a vertex cover of l + 2k nodes. Since there is an edge in every variable gadget, one node in each variable gadget is select. Moreover, two nodes are needed to cover the three edges in each clause gadget. Hence *l* nodes are from the variable gadgets and 2*k* nodes are from the clause gadgets. We assign the *l* literals in the variable gadgets to 1. For each clause gadget, only two nodes are in the vertex cover. Observe that each node in a clause gadget has three edges. To cover the three edges connected to the node not in the vertex cover, the literal corresponding to the node must be 1.

# 3COLOR

• A <u>coloring</u> of a graph is an assignment of colors to its nodes so that adjacent nodes have different colors. Define

 $3COLOR = \{ \langle G \rangle : G \text{ has a coloring with 3 colors} \}.$ 



Theorem 33 3COLOR is NP-complete.

#### Proof.

We reduce  $\neq$  *SAT* to 3*COLOR*. Let *r* be a node. A variable gadget for *x* has two nodes (labeled *x* and  $\overline{x}$ ) with edges {*x*, $\overline{x}$ }, {*r*,*x*}, and {*r*, $\overline{x}$ }. A clause gadget for  $l_0 \lor l_1 \lor l_2$  has three nodes labeled  $l_0, l_1, l_2$  with edges { $l_0, l_1$ }, { $l_0, l_2$ }, { $l_1, l_2$ }. Moreover, each  $l_i$  is connected to the corresponding node in variable gadgets.

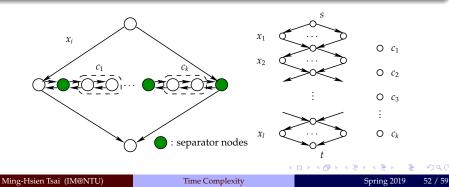
# HAMPATH is NP-Complete

#### Theorem 34

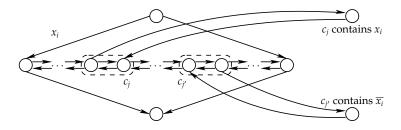
HAMPATH is NP-complete.

### Proof.

We show  $3SAT \leq_P HAMPATH$ . Let  $\phi$  be a 3CNF-formula with l variables and k clauses. We want to construct a graph G with two nodes s and t such that  $\phi$  is satisfiable if and only G has a Hamiltonian path from s to t.



# HAMPATH is NP-Complete



### Proof (cont'd).

Each variable gadget has 4 + 3k nodes. Note that a Hamltonian path traverses a variable gadget in one of the two ways: from left to right, or from right to left. Each clause gadget has 1 node. If  $x_i$  appears in  $c_j$ , two edges are added to visit the clause gadget  $c_j$  from left to right. If  $\overline{x_i}$  appears in  $c_{j'}$ , two edges are added to visit the clause gadget  $c_{j'}$  from right to left.

We show that  $\phi$  is satisfiable if and only if there is a Hamiltonian path from *s* to *t* in *G*. Suppose  $\phi$  has a satisfying assignment. We start from *s* and go through each variable gadget in turn. If  $x_i$  is assigned to 1, we traverse  $x_i$ 's gadget from left to right. If  $x_i$  is assigned to 0, we traverse  $x_i$ 's gadget from right to left. For each clause, one of its literal is 1. We detour to its gadget when traversing variable gadgets.

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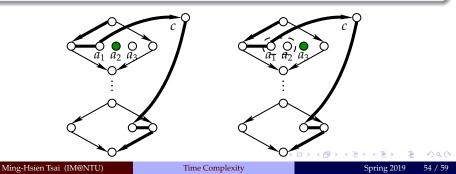
Time Complexity

# HAMPATH is NP-Complete

## Proof (cont'd).

Conversely, suppose *G* has a Hamiltonian path from *s* to *t*. We argue that the Hamiltonian path must traverse every variable gadgets and detour to every clause gadget on its way. Suppose not. Then  $a_2$  or  $a_3$  is a separator node.

- If  $a_2$  is a separator node, it is not connected to any clause gadget. The path must visit  $a_2$  from  $a_3$  since  $a_1$  has been visited. But the path would end in  $a_2$ .
- If *a*<sub>3</sub> is a separator node, *a*<sub>1</sub> and *a*<sub>2</sub> are in the same clause pair. *a*<sub>2</sub> must be visited from *a*<sub>3</sub> since *a*<sub>1</sub> and *c* have been visited. But the path would end in *a*<sub>2</sub> as well. □



# UHAMPATH is NP-Complete

#### Consider

 $UHAMPATH = \{ \langle G, s, t \rangle : G \text{ is an undirected graph with a Hamiltonian path from s to } t \}.$ 

Theorem 35

UHAMPATH is NP-complete.

### Proof.

We show *HAMPATH*  $\leq_P$  *UHAMPATH*. Let *G* be a directed graph. We construct an undirected graph *G'* as follows.

- For each node *u* in *G*, we have three nodes  $u^{\text{in}}, u^{\text{mid}}$ , and  $u^{\text{out}}$  in *G'*. Moreover, *G'* has edges  $\{u^{\text{in}}, u^{\text{mid}}\}$  and  $\{u^{\text{mid}}, u^{\text{out}}\}$ .
- For each edge (u, v) in *G*, add the edge  $\{u^{out}, v^{in}\}$  to *G'*.

We now show that *G* has a Hamiltonian path from *s* to *t* if and only if *G'* has a Hamiltonian path from  $s^{in}$  to  $t^{out}$ .

#### Proof.

For a Hamiltonian path  $s, u_1, u_2, \ldots, u_k, t$  in  $G, s^{in}, s^{mid}, s^{out}, u_1^{in}, u_1^{mid}, u_2^{out}, u_2^{in}, \ldots, u_k^{in}, u_k^{mid}, u_k^{out}, t^{in}, t^{mid}, t^{out}$  is a Hamiltonian path in G'. Conversely, any Hamiltonian path in G' from  $s^{in}$  must be of the form

 $s^{\text{in}}, s^{\text{mid}}, s^{\text{out}}, u_i^{\text{in}}, \dots, (\text{for some } i).$ 

But this corresponds to the edge  $(s, u_i)$  in *G*. Moreover, the next node must be  $u_i^{\text{mid}}$ , and then followed by  $u_i^{\text{out}}$ . The Hamiltonian path now looks like

 $s^{\text{in}}, s^{\text{mid}}, s^{\text{out}}, u_i^{\text{in}}, u_i^{\text{mid}}, u_i^{\text{out}}, u_j^{\text{in}}, \dots, (\text{for some } j).$ 

But this corresponds to the edges  $(s, u_i)$  and  $(u_i, u_j)$  in *G*. Eventually we reach  $t^{\text{out}}$  and obtain a Hamiltonian path from *s* to *t* in *G*.

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# SUBSETSUM is NP-Complete

#### Theorem 36

SUBSETSUM is NP-complete.

#### Proof.

We show  $3SAT \leq_P SUBSETSUM$ . Let  $\phi$  be a 3CNF-formula with l variables and k clauses. We construct a multiset S so that  $\phi$  is satisfiable if and only if the sum of a

subset of *S* is  $t = \overbrace{1 \cdots 1}^{\infty} 3$  (a very large decimal number).

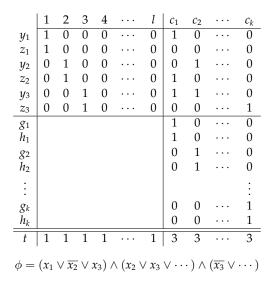
• For each variable *x<sub>i</sub>*, we add numbers *y<sub>i</sub>* and *z<sub>i</sub>* to *S* if *x<sub>i</sub>* appears in *c<sub>r</sub>*, *c<sub>t</sub>* and *x<sub>i</sub>* appears in *c<sub>s</sub>* (unspecified digits are 0):

• For each clause  $c_j$ , add  $g_j$  and  $h_j$  to S (unspecified digits are 0):

$$g_j = h_j = \left| \begin{array}{ccc} \cdots & \cdots & c_j & \cdots \\ & & 1 \end{array} \right|$$

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## SUBSETSUM is NP-Complete



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### Proof.

Suppose  $\phi$  has a satisfying truth assignment. We put  $y_i$  to the subset if  $x_i$  is assigned to 1; we put  $z_i$  if  $x_i$  is assigned to 0. Since either  $y_i$  or  $z_i$  is selected for each i, the first l digits of the sum are all 1's. For each of the last k digits of the sum, it cannot be 0 since one of the literals in the clause is 1. We add  $g_i$ 's or  $h_i$ 's to the subset so that the last k digits of the sum are all 3's.

Conversely, suppose *S* has a subset whose sum is *t*. Observe that each digit has at most 5 1's. There cannot be any carry. Either  $y_i$  or  $z_i$  must be selected for each *i* since the first *l* digits are all 1's. This gives us a truth assignment for  $\phi$ . Finally, note that the last *k* digits are all 3's. Yet  $g_j$  and  $h_j$  contribute at most 2 for each of the last *k* digits. Subsequently,  $y_i$ 's or  $z_i$ 's contribute at least 1 for each of the last *k* digits. The truth assignment does satisfy  $\phi$ .

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