



# Computing Theory

COMP 147 (4 units)

Chapter 7: Time Complexity

Class NP

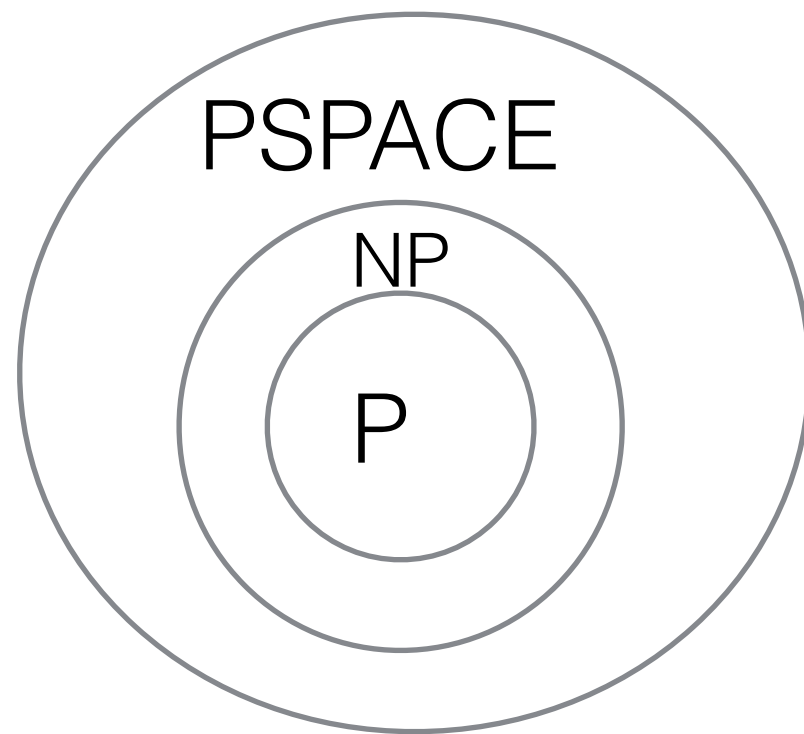
NP Completeness

# Last Time

- Section 7.1: Time Complexity
- Section 7.2: Class P

$$P = \bigcup_k \text{TIME}(n^k)$$

# Complexity Classes



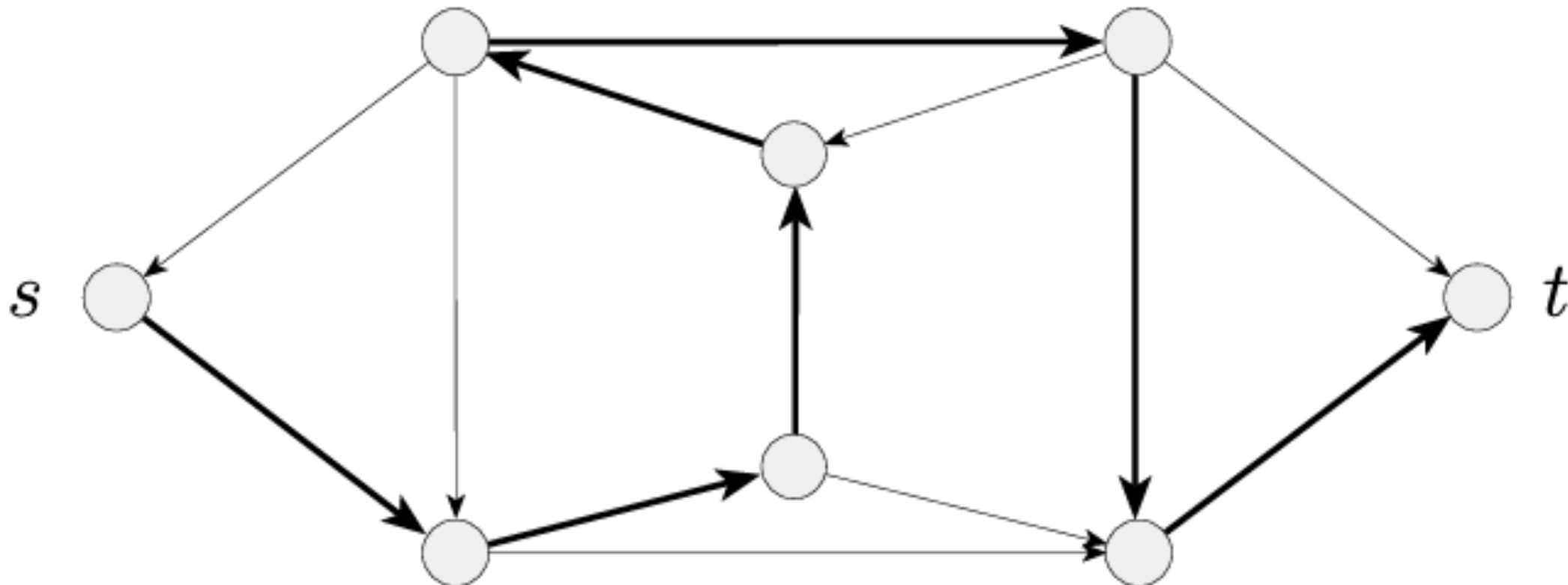
NP: nondeterministic polynomial time

# Decision problems

- Problem
  - Sort this list of numbers
  - Find the shortest path between  $s$  and  $t$  in a graph
- Decision Version
  - Is this list sorted?
  - Is there a shorter path less than  $k$

# ***HAMPATH*** $\notin P$ ?

- $HAMPATH = \{ \langle G, s, t \rangle \mid G \text{ is a directed graph with a Hamiltonian path from } s \text{ to } t \}$
- A Hamiltonian path visits every node exactly once.



# ***HAMPATH***: Brute force algorithm

- On input  $\langle G, s, t \rangle$ :
  - Generate a permutation of all nodes of  $G$ , call these sequence  $s_1, s_2, \dots, s_{n!}$
  - For  $i = 1$  to  $n!$ :
    - If  $s_i$  is a valid path from  $s$  to  $t$ , accept
  - Reject (all sequence failed test)
- This algorithm is  $O(n!)$
- There is no known algorithm for  $\text{HAMPATH} \in P$

# Verifiers

## DEFINITION 7.18

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

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

We measure the time of a verifier only in terms of the length of  $w$ , so a *polynomial time verifier* runs in polynomial time in the length of  $w$ . A language  $A$  is *polynomially verifiable* if it has a polynomial time verifier.

- A verifier only needs to check if a solution is valid, it does not need to generate a solution

# HAMPATH is polynomially verifiable

- $V_{HAMPATH} = \{ \langle G, s, t \rangle, c \mid G \text{ is a directed graph and } c \text{ is a Hamiltonian path from } s \text{ to } t \}$ 
  - Let  $c = c_1, c_1, \dots, c_n$
  - Verify that  $c_1 = s$
  - Verify that  $c_n = t$
  - For  $i = 1$  to  $n-1$ 
    - Verify that an edge exists from  $c_i$  to  $c_{i+1}$
  - If all tests pass, accept, otherwise reject
- Complexity?

in polynomial time



# Verifiers and Certificates

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

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

- $c$  is called a certificate
  - It certifies that a solution to the problem exists
  - Usually,  $c$  is simply a solution to the problem

# The Class NP

- Definition 7.19:  
NP is the class of languages that have polynomial time verifiers.
- (True or False) If  $A \in P$ , then  $A \in NP$
- $HAMPATH \in NP$ 
  - $HAMPATH \in P$ ?

**P** = The class of languages  
for which membership can  
be **DECIDED** quickly.\*

**NP** = The class of languages  
for which membership can  
be **VERIFIED** quickly.

↖ That is, given some information  
[the "certificate/proof"], you can  
quickly confirm that  $w$  is  
in the language.

# NTIME

## DEFINITION 7.21

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$\text{NTIME}(t(n)) = \{L \mid L \text{ is a language decided by an } O(t(n)) \text{ time nondeterministic Turing machine}\}.$

- Recall that the running time of an NTM is the length of the longest branch in the computation history

# The Class NP

- Theorem 7.20  
A language is in NP iif it is decided by some nondeterministic polynomial time TM.

**COROLLARY 7.22** ....

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



### DEFINITION

"NP" is the class of languages that have polynomial-time verifiers.

### THEOREM

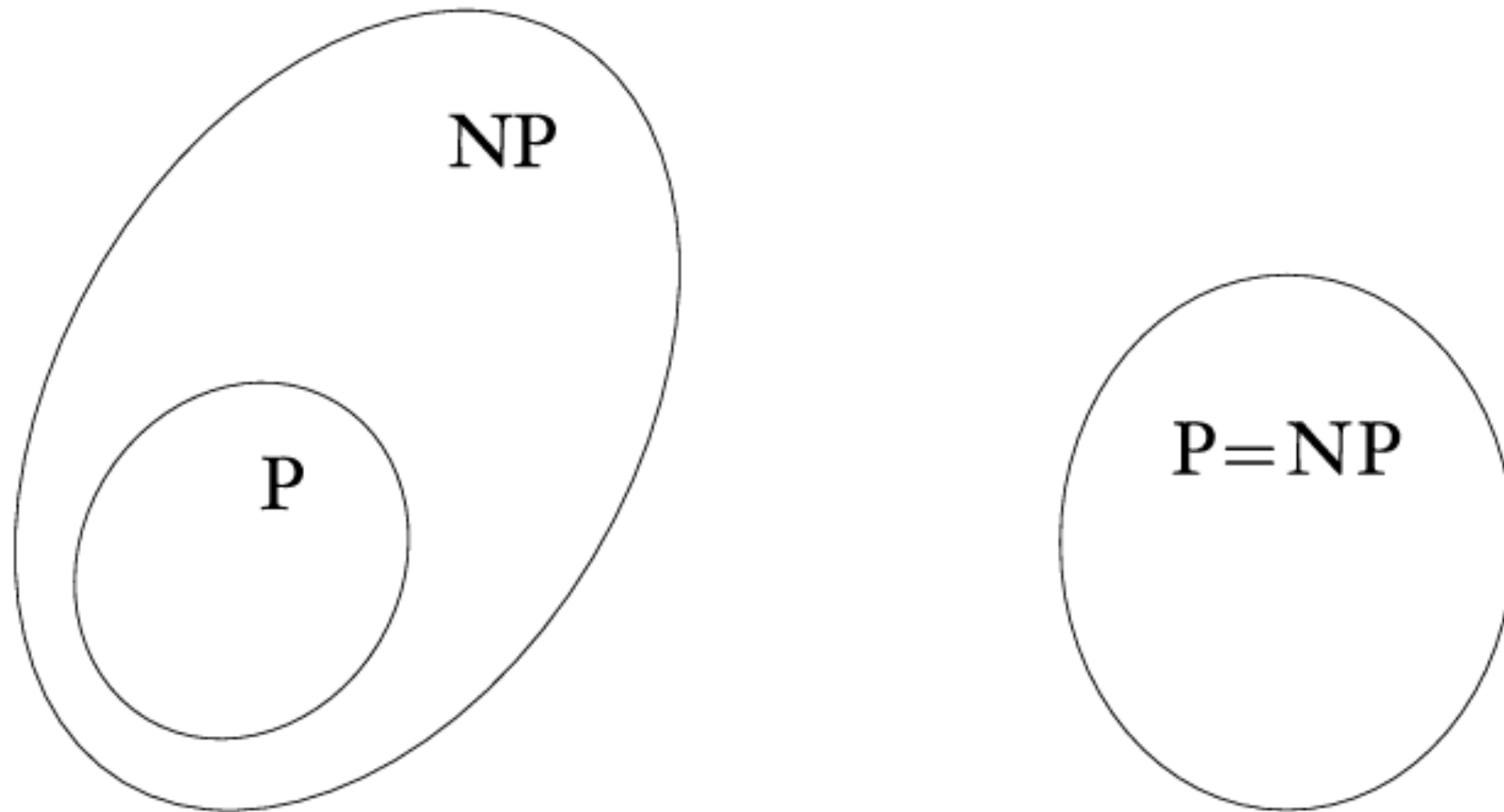
A language is in NP iff it is decided by some NONDETERMINISTIC POLYNOMIAL-TIME Turing Machine

Sometimes this is given as the definition of "NP".

# $P = NP?$

UNSOLVED QUESTION:

$P = NP$   
 $P \subset NP$  } Which is it?



**FIGURE 7.26**

One of these two possibilities is correct

# Best NP Bound (so far)

- The best deterministic method currently known for deciding languages in NP uses exponential time.
- We can prove that

$$\text{NP} \subseteq \text{EXPTIME} = \bigcup_k \text{TIME}(2^{n^k})$$

- We do not know if NP is contained in a smaller deterministic time complexity class.

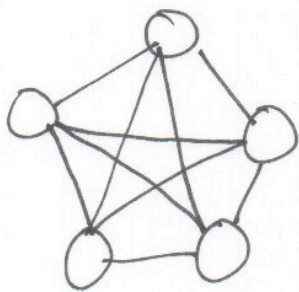


# The Clique Problem

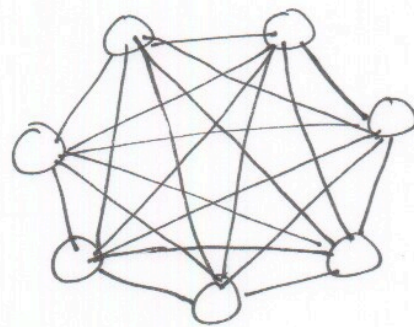
Given an undirected graph...

A "clique" is a set of nodes such that every node in the clique is connected to every other node in the clique.

A  $K$ -clique is a clique with  $K$  members.



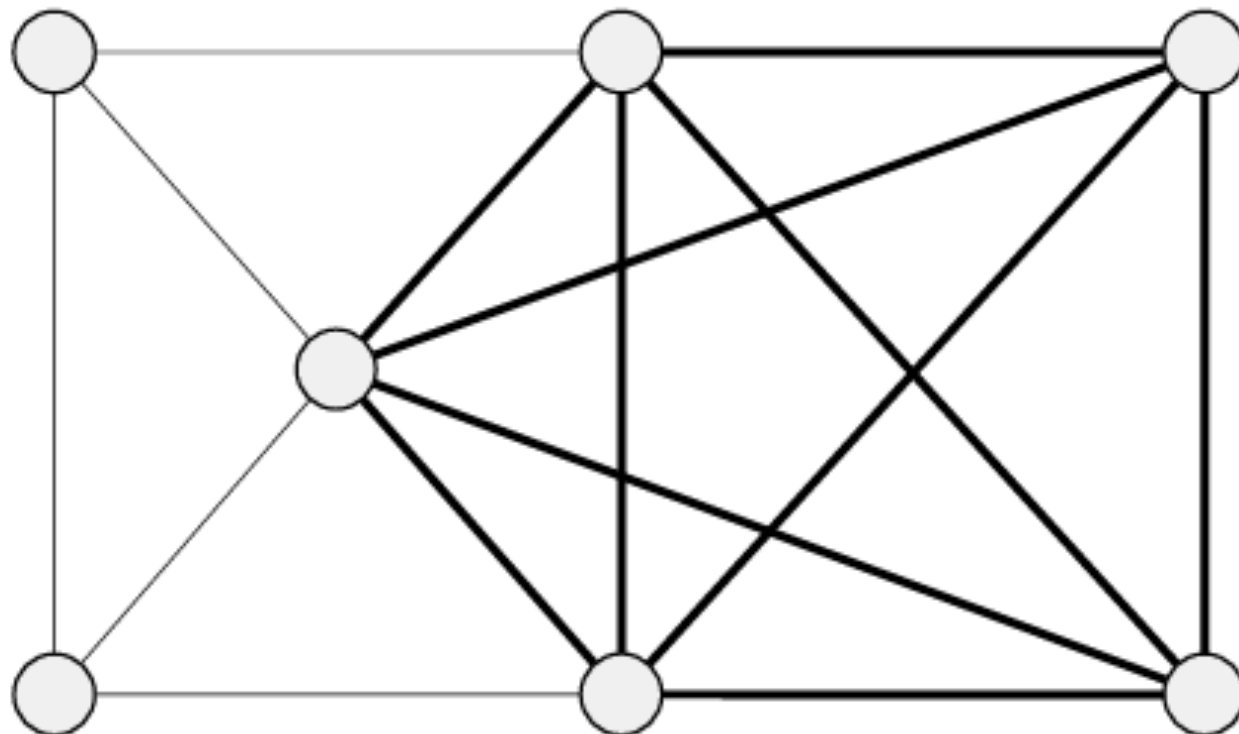
A 5-CLIQUE



A 7-CLIQUE.

# ***CLIQUE***

- $CLIQUE = \{ \langle G, k \rangle \mid G \text{ is an undirected graph with a } k\text{-clique} \}$ 
  - clique = fully connected subgraph
  - $k$ -clique = fully connected subgraph of  $k$  nodes



Graph with a 5-clique

# *CLIQUE* $\in$ NP

Using definition 1: Polynomial time verifier

**THEOREM 7.24** .....

*CLIQUE* is in NP.

**PROOF IDEA** The clique is the certificate.

**PROOF** The following is a verifier  $V$  for *CLIQUE*.

$V =$  “On input  $\langle\langle G, k \rangle, c\rangle$ :

1. Test whether  $c$  is a subgraph with  $k$  nodes in  $G$ .
2. Test whether  $G$  contains all edges connecting nodes in  $c$ .
3. If both pass, *accept*; otherwise, *reject*.”

# *CLIQUE* $\in$ NP

Using definition2: Nondeterministic Turing Machines

**THEOREM 7.24** .....

*CLIQUE* is in NP.

**ALTERNATIVE PROOF** If you prefer to think of NP in terms of nondeterministic polynomial time Turing machines, you may prove this theorem by giving one that decides *CLIQUE*. Observe the similarity between the two proofs.

$N =$  “On input  $\langle G, k \rangle$ , where  $G$  is a graph:

1. Nondeterministically select a subset  $c$  of  $k$  nodes of  $G$ .
2. Test whether  $G$  contains all edges connecting nodes in  $c$ .
3. If yes, *accept*; otherwise, *reject*.”

# SAT $\in$ NP

**Example:** Determine the satisfiability of the following compound propositions:

$$(p \vee \neg q) \wedge (q \vee \neg r) \wedge (r \vee \neg p)$$

**Solution:** Satisfiable. Assign T to p, q, and r.

$$(p \vee q \vee r) \wedge (\neg p \vee \neg q \vee \neg r)$$

**Solution:** Satisfiable. Assign T to p and F to q.

$$(p \vee \neg q) \wedge (q \vee \neg r) \wedge (r \vee \neg p) \wedge (p \vee q \vee r) \wedge (\neg p \vee \neg q \vee \neg r)$$

**Solution:** Not satisfiable. Check each possible assignment of truth values to the propositional variables and none will make the proposition true.

# NP-Completeness

# Complete Problems

- One way to address the  $P = NP$  question is to identify **complete problems** for NP.
- An **NP-complete problem** has the property that it is in **NP**, and if it is in **P**, then every problem in **NP** is also in **P**.
- Defined formally via “polytime reductions.”

# NP-Completeness

- A problem  $L$  is NP-complete if
  - 1.  $L \in \text{NP}$ , and
  - 2. Every problem  $L' \in \text{NP}$ ,  
 $L'$  is polytime reducible to  $L$  in polynomial time
- $L$  is as hard as any problem in NP



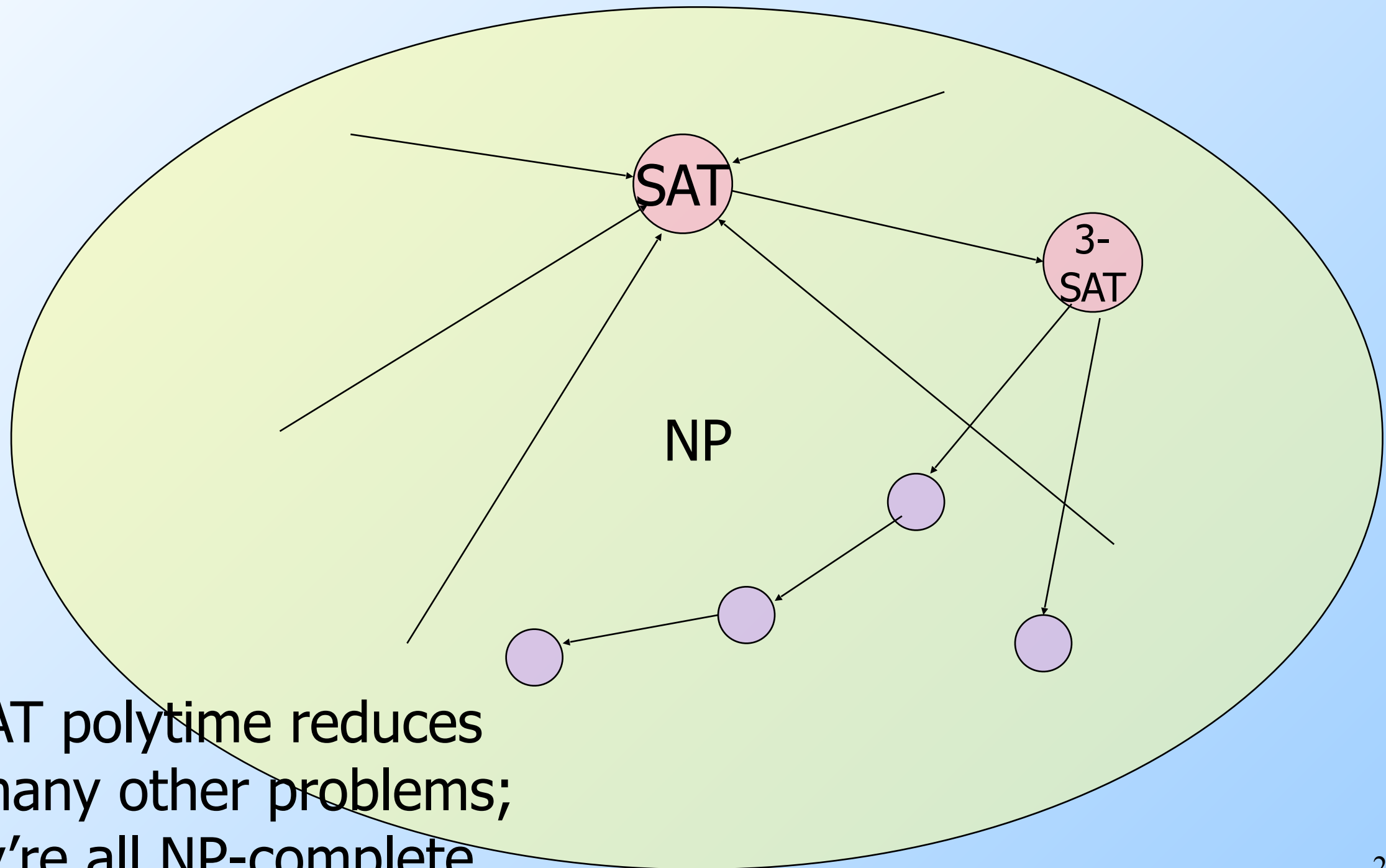
- If a polynomial time algorithm is ever found (on a deterministic machine) for any "NP-Complete" problem, then  $P=NP$  follows!

... And polynomial time algorithms exist for all problems in NP!

All of **NP** polytime  
reduces to SAT, which  
is therefore NP-complete

# The Plan

SAT polytime  
reduces to  
3-SAT



3-SAT polytime reduces  
to many other problems;  
they're all NP-complete

# Satisfiability (*SAT*)

- $SAT = \{ \langle \phi \rangle \mid \phi \text{ is a satisfiable Boolean formula} \}$
- Suppose the formula has  $m$  variables and  $n$  operations.
  - How many possible assignments?  $2^m$
- Given an assignment, what is the complexity of verification?  $O(n)$

# 3SAT

- $3SAT = \{ \langle \varphi \rangle \mid \varphi \text{ is a satisfiable 3cnf formula} \}$
- literal: a Boolean variable or a negated Boolean variable
- clause: literals connected by disjunction ( $\vee$ )
- ■ cnf formula: clauses connected by conjunction ( $\wedge$ )
- 3cnf formula: each clause has exactly 3 literals

$$(x_1 \vee \overline{x_2} \vee \overline{x_3}) \wedge (x_3 \vee \overline{x_5} \vee x_6) \wedge (x_3 \vee \overline{x_6} \vee x_4) \wedge (x_4 \vee x_5 \vee x_6).$$

# 3SAT

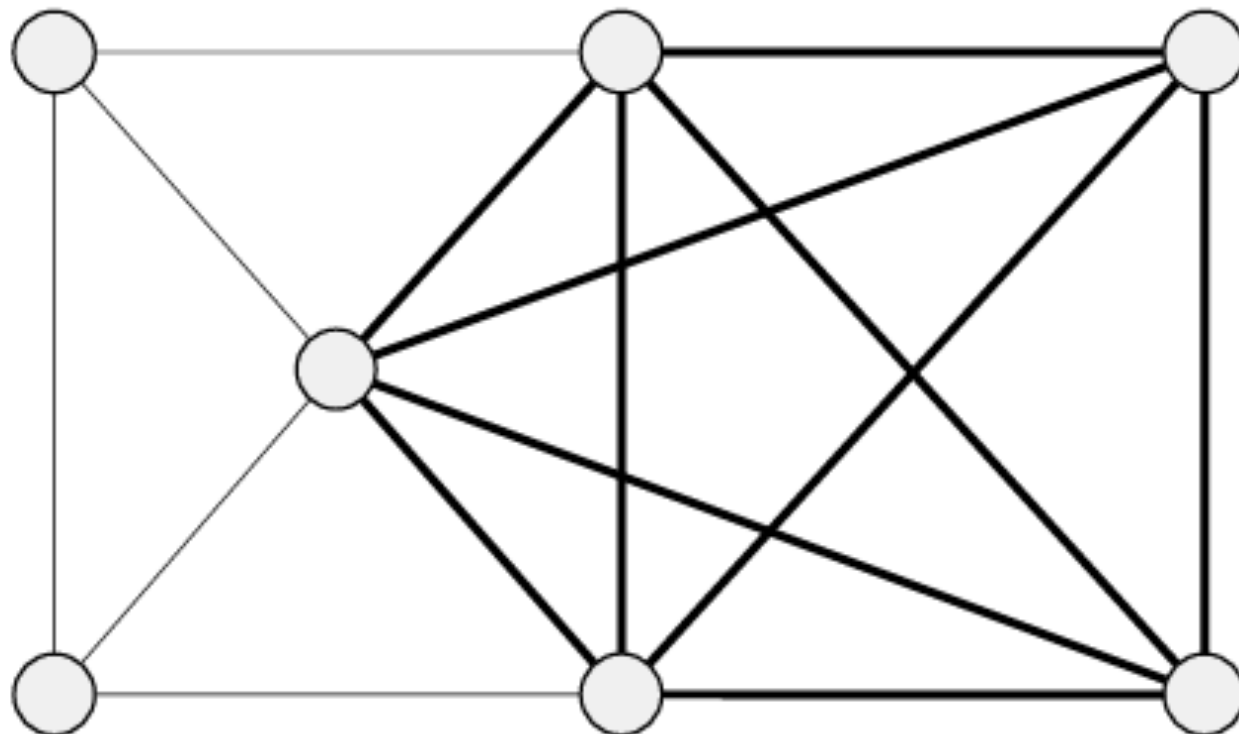
- $3SAT = \{ \langle \phi \rangle \mid \phi \text{ is a satisfiable 3cnf formula} \}$

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- Each clause must have at least one true literal
- Example has 64 possible assignments

# CLIQUE

- $CLIQUE = \{ \langle G, k \rangle \mid G \text{ is an undirected graph with a } k\text{-clique} \}$ 
  - clique = fully connected subgraph
  - $k$ -clique = fully connected subgraph of  $k$  nodes

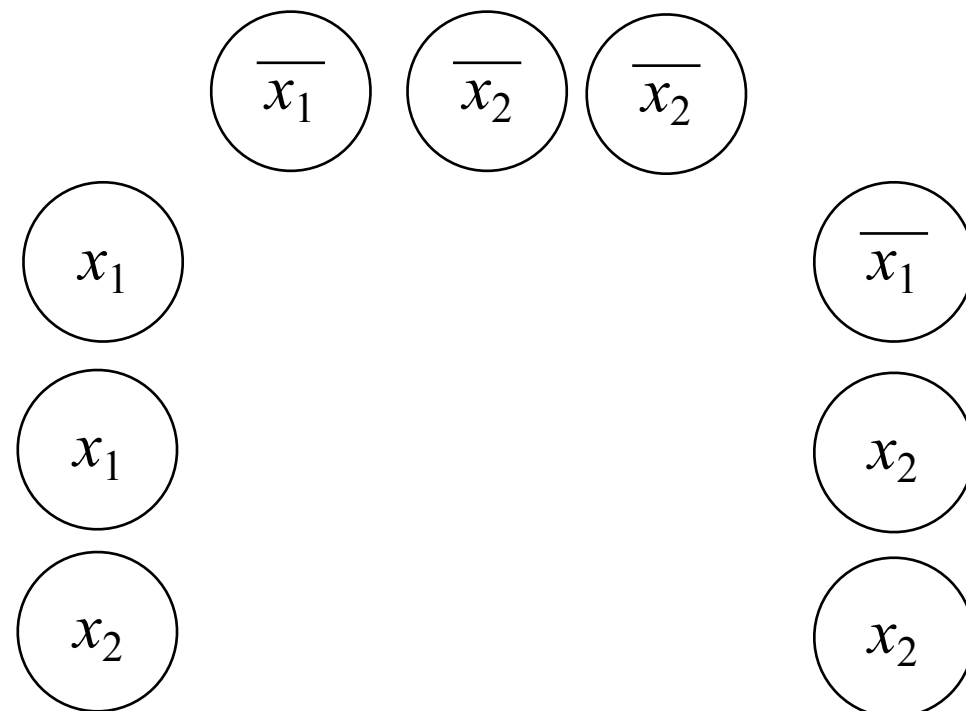


Graph with a 5-clique

# $3SAT \leq_P CLIQUE$

- Given a 3cnf-formula with  $k$  clauses, generate a graph with  $3k$  nodes, such that the formula is satisfiable iff the graph has a  $k$ -clique.

$$\phi = (x_1 \vee x_1 \vee x_2) \wedge (\overline{x_1} \vee \overline{x_2} \vee \overline{x_2}) \wedge (\overline{x_1} \vee x_2 \vee x_2)$$

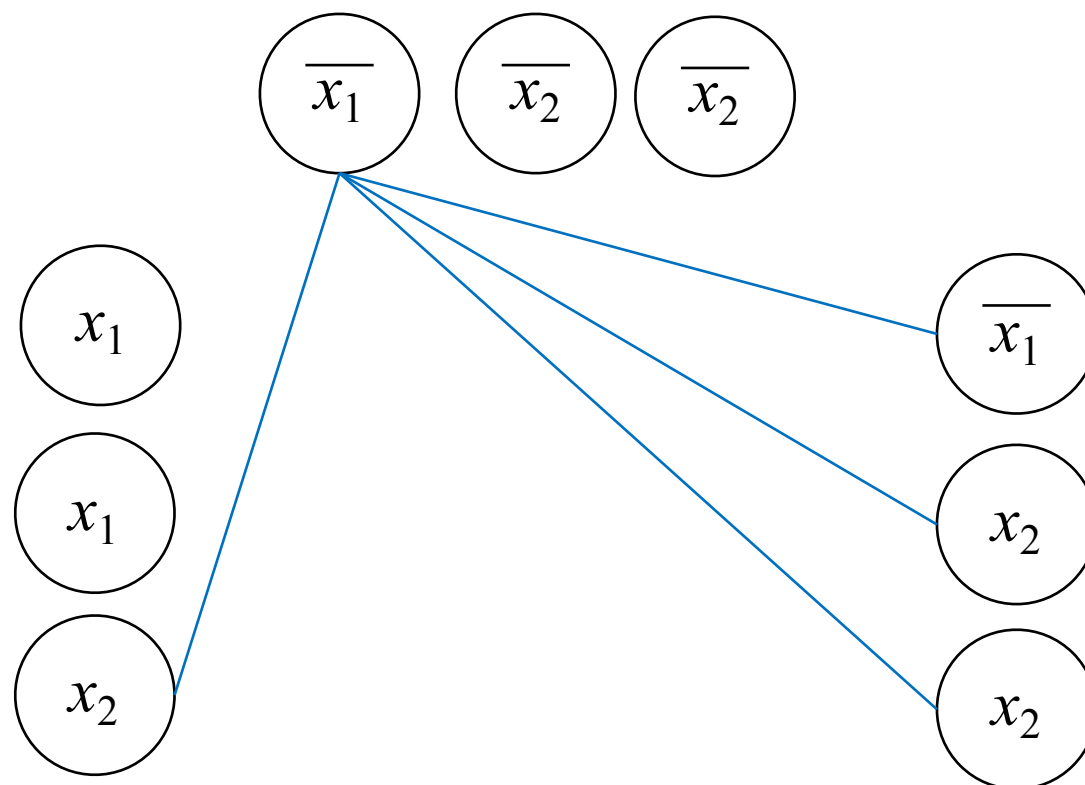




# $3SAT \leq_P CLIQUE$

- Connect each pair of nodes, unless:
  - they are in the same clause, or
  - they have contradictory (negated) labels

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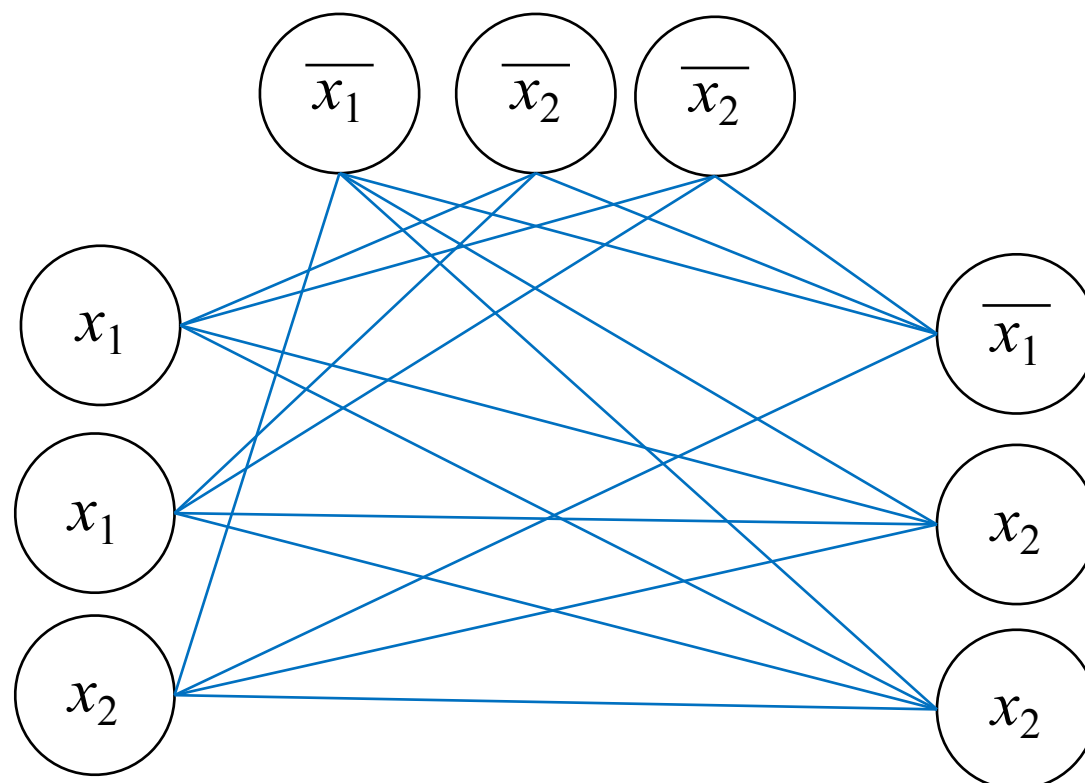




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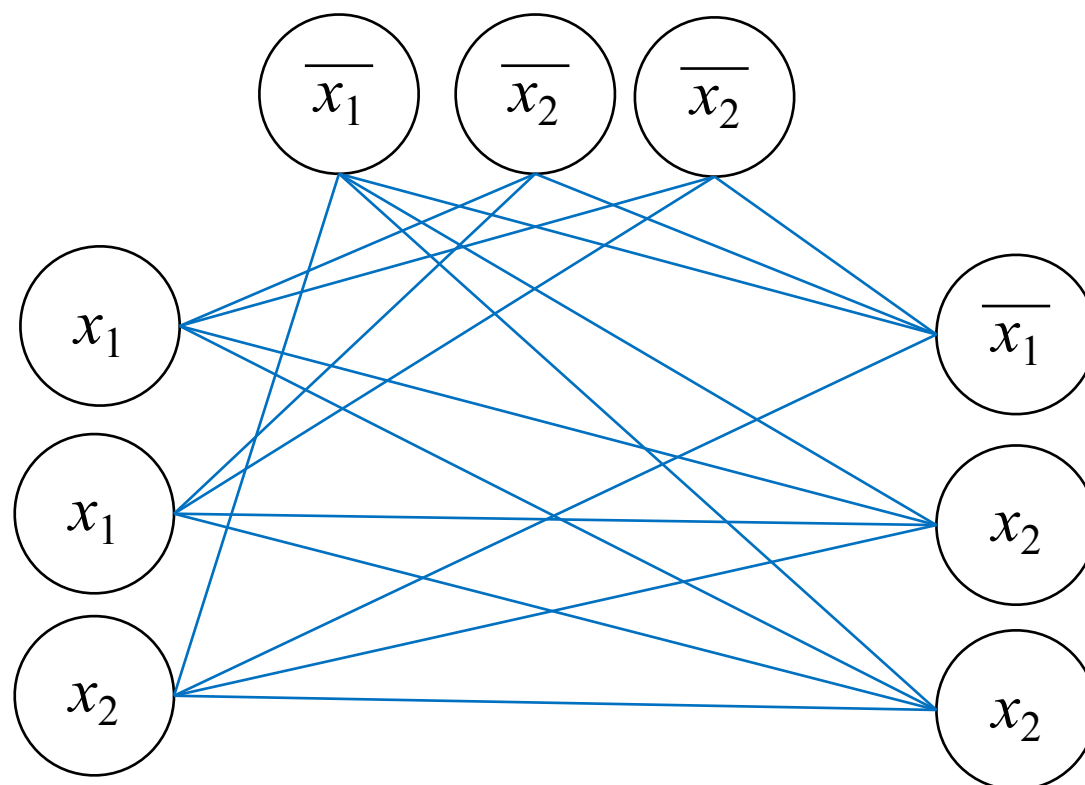
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# $3SAT \leq_P CLIQUE$

- Any  $k$ -clique
  - has at most one node from each clause  $\rightarrow$   
has exactly one node from each clause
  - Does not contain contradictory assignments

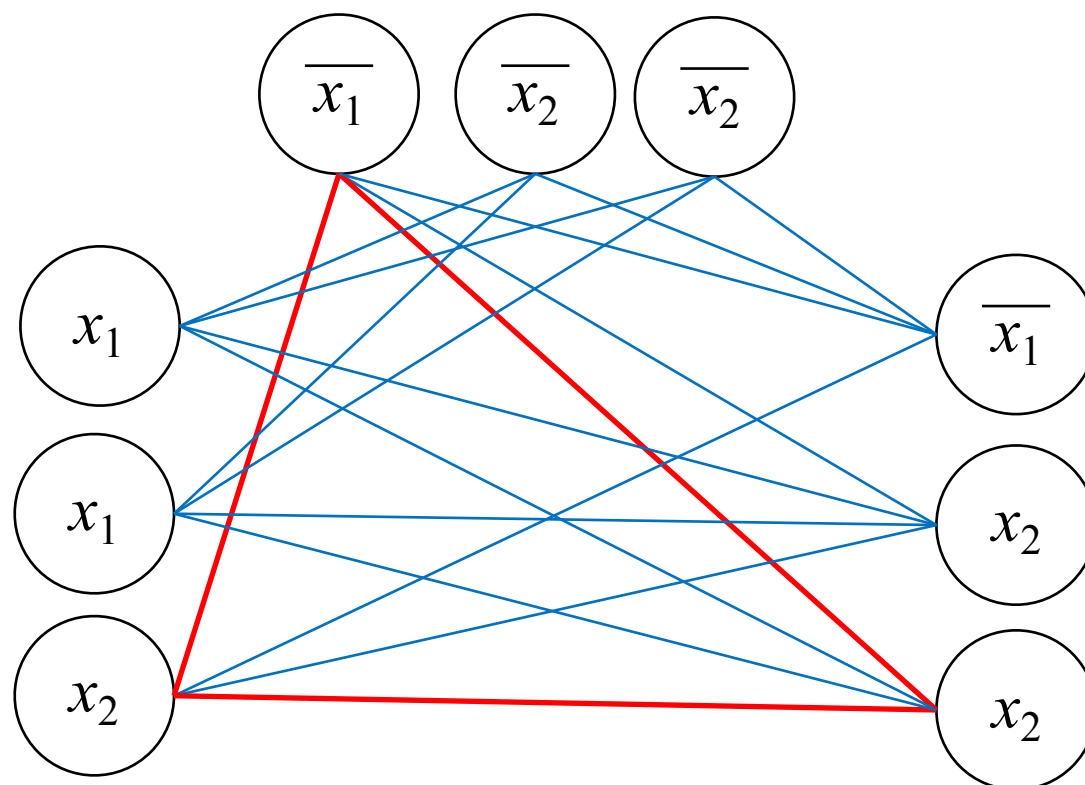
$$\phi = (x_1 \vee x_1 \vee x_2) \wedge (\overline{x_1} \vee \overline{x_2} \vee \overline{x_2}) \wedge (\overline{x_1} \vee x_2 \vee x_2)$$



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# Initial NP-complete Language

- Theorem 7.37: ***SAT*** is NP-complete

- Proof Idea:

- 1. Show that  $SAT \in NP$  (easy)
- 2. Show that any language  $A \in NP$  is ptime-reducible to  $SAT$

Given  $\langle A, w \rangle$  we'll construct a Boolean formula  $\phi$  that simulates the NP machine  $M$  for  $A$  on  $w$ .

$M$  accepts  $w \Leftrightarrow \phi$  is satisfiable

$M$  doesn't accept  $w \Leftrightarrow \phi$  is not satisfiable

Given that AND, OR and NOT are the basic components of digital computers, it's not surprising that we can simulate a TM with a logical formula. However, the devil is in the details.

# *SAT* $\in$ NP

- A nondeterministic poly-time TM can guess an assignment for a given formula  $\phi$  and accept if the assignment satisfies  $\phi$ .
- Clearly, verification is  $O(n^k)$ , therefore SAT  $\in$  NP