

Intro to Theory of Computation

CS
332

LECTURE 23

Last time

- Dynamic programming proof that all CFLs are in P
- Class NP

Today

- The P vs. NP question
- Polynomial-time reductions
- NP-completeness

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Classes P, NP, EXP

- **P.** Class of languages for which there is a **poly-time algorithm**.
algorithm that runs in time $O(n^k)$ for some k
- **EXP.** Class of languages for which there is an **exponential-time algorithm**.
algorithm that runs in time $O(2^{n^k})$ for some k
- **NP.** Class of languages for which there is a **poly-time verifier**.
- **Lemma.** $P \subseteq NP$.
- **Lemma.** $NP \subseteq EXP$.
- **Lemma.** A language L is in NP iff L can be decided by a polynomial-time nondeterministic TM.

To prove $\text{NP} \subseteq \text{EXP}$, consider a language $L \in \text{NP}$.

Then L has a verifier V that runs in time n^k .

We can construct an $O(2^{n^k})$ -time TM for L as follows:

- A. “On input $\langle w, c \rangle$, where c is a certificate, run V on $\langle w, c \rangle$.”
- B. “On input w , run V on $\langle w, c \rangle$ for all possible certificates c .”
- C. “On input w , run V on $\langle w, c \rangle$ for all possible certificates c of length at most $|w|^k$.”
- D. “On input w , run L on $\langle w, c \rangle$ for all possible certificates c of length at most $n^{|w|}$.”
- E. None of the above.

Nondeterministic time complexity classes

$\text{NTIME}(f(n))$ is a class of languages.

$A \in \text{NTIME}(f(n))$ means that

some nondeterministic TM M

that runs in time $O(f(n))$ decides A .

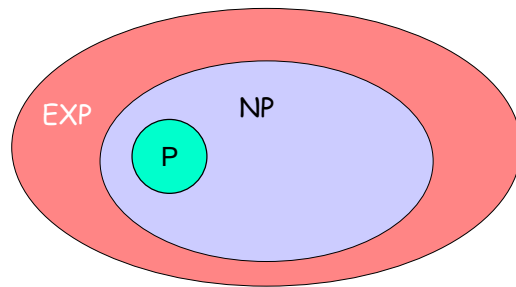
The class NP: alternative definition

NP is the class of languages decidable in polynomial time on a *nondeterministic* TM:

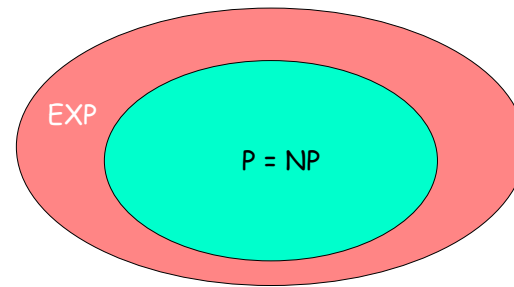
$$\mathbf{NP} = \bigcup_k \mathit{NTIME}(n^k).$$

P vs. NP

- Does $P = NP$? [Cook 1971, Edmonds, Levin, Yablonski, Gödel]
 - Is the decision problem as easy as the verification problem?
 - Clay \$1 million prize.



If $P \neq NP$



If $P = NP$ would break RSA cryptography
(and potentially collapse economy) ↘

- If yes: Efficient algorithms for UHamPath, SAT, TSP, factoring
- If no: No efficient algorithms possible for these problems.
- Consensus opinion on $P = NP$? Probably no.

Classifying Problems

- **Desiderata:** classify problems according to those that can be solved in polynomial-time and those that cannot.
- Some problems *provably require exponential time* (Chapter 9):
 - Given a Turing machine, does it halt in at most k steps?
 - Given a board position in an n -by- n generalization of chess, can black guarantee a win?
- **Frustrating news:** huge number of fundamental problems have defied classification for decades.
- **Chapters 7.4-7.5 (NP-completeness):** Show that these fundamental problems are "computationally equivalent" and appear to be different manifestations of one **really hard** problem.

Polynomial-time reduction

Given languages A and B,

$$A \leq_p B$$

if there is a *poly-time* computable function f ,
such that for all strings w ,
 $w \in A$ iff $f(w) \in B$.



Polynomial-time reductions
are the major tool we have
to understand P and NP

Implication of poly-time reductions

Theorem. If $A \leq_p B$ and $B \in \mathbf{P}$ then $A \in \mathbf{P}$.

(So, if $A \leq_p B$ and $A \notin \mathbf{P}$ then $B \notin \mathbf{P}$.)

Theorem. If $A \leq_p B$ and $B \leq_p C$ then $A \leq_p C$.

(Poly-time reductions compose.)

Basic reduction strategies

Basic reduction strategies

- Reduction by simple equivalence.
- Reduction from special case to general case.
- Reduction by encoding with gadgets.

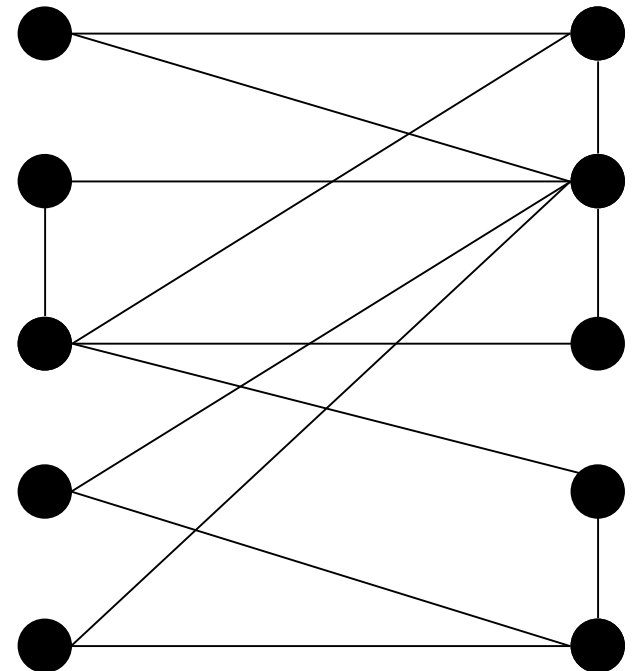
Independent Set

Given an undirected graph G , an **independent set** in G is a set of nodes, which includes at most one endpoint of every edge.

INDEPENDENT SET = $\{\langle G, k \rangle \mid G \text{ is an undirected graph which has an independent set with } k \text{ nodes}\}$

- Is there an independent set of size ≥ 6 ?
 - Yes.
- Is there an independent set of size ≥ 7 ?
 - No.

● independent set



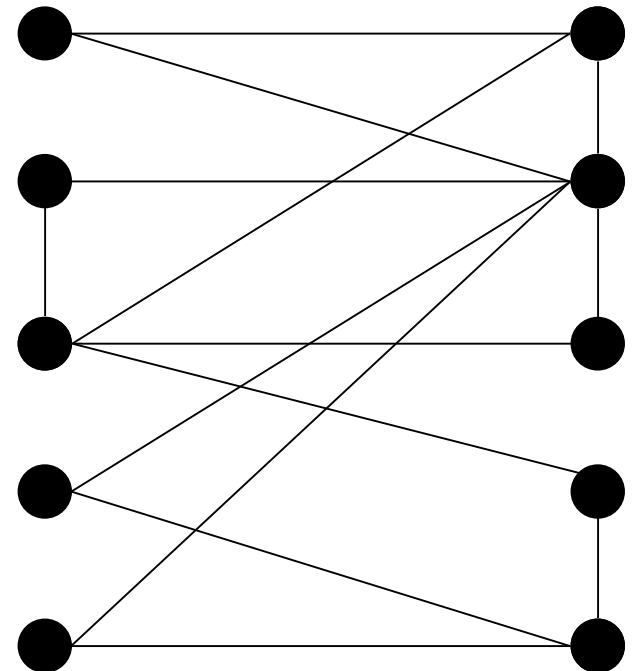
Vertex Cover

Given an undirected graph G , a **vertex cover** in G is a set of nodes, which includes at *least* one endpoint of every edge.

VERTEX COVER = $\{\langle G, k \rangle \mid G \text{ is an undirected graph which has a vertex cover with } k \text{ nodes}\}$

- Is there vertex cover of size ≤ 4 ?
 - Yes.
- Is there a vertex cover of size ≤ 3 ?
 - No.

● vertex cover



Independent Set and Vertex Cover

Claim. S is an independent set iff $V - S$ is a vertex cover.

- \Rightarrow
 - Let S be any independent set.
 - Consider an arbitrary edge (u, v) .
 - S is independent $\Rightarrow u \notin S$ or $v \notin S \Rightarrow u \in V - S$ or $v \in V - S$.
 - Thus, $V - S$ covers (u, v) .
- \Leftarrow
 - Let $V - S$ be any vertex cover.
 - Consider two nodes $u \in S$ and $v \in S$.
 - Then $(u, v) \notin E$ since $V - S$ is a vertex cover.
 - Thus, no two nodes in S are joined by an edge $\Rightarrow S$ independent set. ■

INDEPENDENT SET reduces to VERTEX COVER

Theorem. $\text{INDEPENDENT-SET} \leq_p \text{VERTEX-COVER}$.

Proof. “On input $\langle G, k \rangle$, where G is an undirected graph and k is an integer,

1. Output $\langle G, n - k \rangle$, where n is the number of nodes in G .”

Correctness:

- G has an independent set of size k iff it has a vertex cover of size $n - k$.
- Reduction runs in linear time.

Reduction from special case to general case

Basic reduction strategies

- Reduction by simple equivalence.
- Reduction from special case to general case.
- Reduction by encoding with gadgets.

Given a set U , called a *universe*, and a collection of its subsets S_1, S_2, \dots, S_m , a **set cover** of U is a subcollection of subsets whose union is U .

- $\text{SET COVER} = \{ \langle U, S_1, S_2, \dots, S_m; k \rangle \mid U \text{ has a set cover of size } k \}$

- Sample application.

- m available pieces of software.
- Set U of n capabilities that we would like our system to have.
- The i th piece of software provides the set $S_i \subseteq U$ of capabilities.
- Goal: achieve all n capabilities using fewest pieces of software.

$U = \{ 1, 2, 3, 4, 5, 6, 7 \}$

$k = 2$

$S_1 = \{3, 7\}$

$S_4 = \{2, 4\}$

$S_2 = \{3, 4, 5, 6\}$

$S_5 = \{5\}$

$S_3 = \{1\}$

$S_6 = \{1, 2, 6, 7\}$

VERTEX COVER reduces to SET COVER

Theorem. $\text{VERTEX-COVER} \leq_p \text{SET-COVER}$.

Proof. “On input $\langle G, k \rangle$, where $G = (V, E)$ is an undirected graph and k is an integer,

1. Output $\langle U, S_1, S_2, \dots, S_m; k \rangle$, where $U=E$ and
$$S_v = \{e \in E \mid e \text{ incident to } v\}$$
”

Correctness:

- G has a vertex cover of size k iff U has a set cover of size k .
- Reduction runs in linear time.

Reduction by encoding with gadgets

Basic reduction strategies

- Reduction by simple equivalence.
- Reduction from special case to general case.
- Reduction by encoding with gadgets.

Satisfiability

- **Boolean variables:** variables that can take on values T/F (or 1/0)
- **Boolean operations:** \vee , \wedge , and \neg
- **Boolean formula:** expression with Boolean variables and ops

$\text{SAT} = \{ \langle \Phi \rangle \mid \Phi \text{ is a satisfiable Boolean formula} \}$

- **Literal:** A Boolean variable or its negation. x_i or $\overline{x_i}$
- **Clause:** OR of literals. $C_j = x_1 \vee \overline{x_2} \vee x_3$
- **Conjunctive normal form (CNF):** AND of clauses. $\Phi = C_1 \wedge C_2 \wedge C_3 \wedge C_4$

$3\text{SAT} = \{ \langle \Phi \rangle \mid \Phi \text{ is a satisfiable Boolean CNF formula, where each clause contains exactly 3 literals} \}$

↑
each corresponds to a different variable

Ex: $(\overline{x_1} \vee x_2 \vee x_3) \wedge (x_1 \vee \overline{x_2} \vee x_3) \wedge (x_2 \vee x_3) \wedge (\overline{x_1} \vee \overline{x_2} \vee \overline{x_3})$

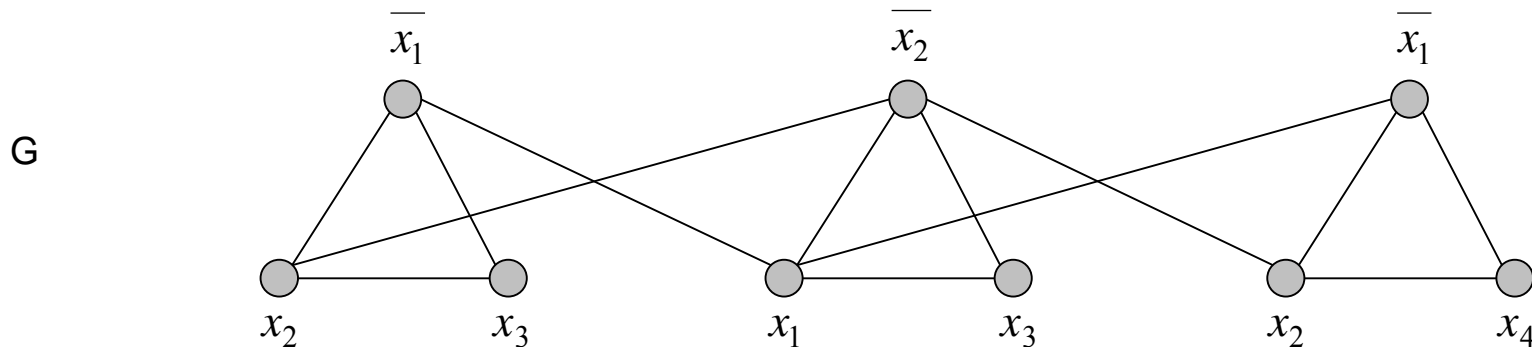
Yes: $x_1 = \text{true}, x_2 = \text{true}, x_3 = \text{false}.$

3SAT reduces to INDEPENDENT SET

Theorem. $3\text{-SAT} \leq_p \text{INDEPENDENT-SET}$.

Proof. “On input $\langle \Phi \rangle$, where Φ is a 3CNF formula,

1. Construct graph G from Φ
 - G contains 3 vertices for each clause, one for each literal.
 - Connect 3 literals in a clause in a triangle.
 - Connect literal to each of its negations.
2. Output $\langle G, k \rangle$, where k is the number of clauses in G .”



$k = 3$

$$\Phi = (\bar{x}_1 \vee x_2 \vee x_3) \wedge (x_1 \vee \bar{x}_2 \vee x_3) \wedge (\bar{x}_1 \vee x_2 \vee x_4)$$

3SAT reduces to INDEPENDENT SET

Correctness. Let $k = \#$ of clauses and $\ell = \#$ of literals in Φ .

Φ is satisfiable iff G contains an independent set of size k .

- \Rightarrow Given satisfying assignment, select one true literal from each triangle. This is an independent set of size k .
- \Leftarrow Let S be an independent set of size k .
 - S must contain exactly one vertex in each triangle.
 - Set these literals to true, and other literals in a consistent way.
 - Truth assignment is consistent and all clauses are satisfied.

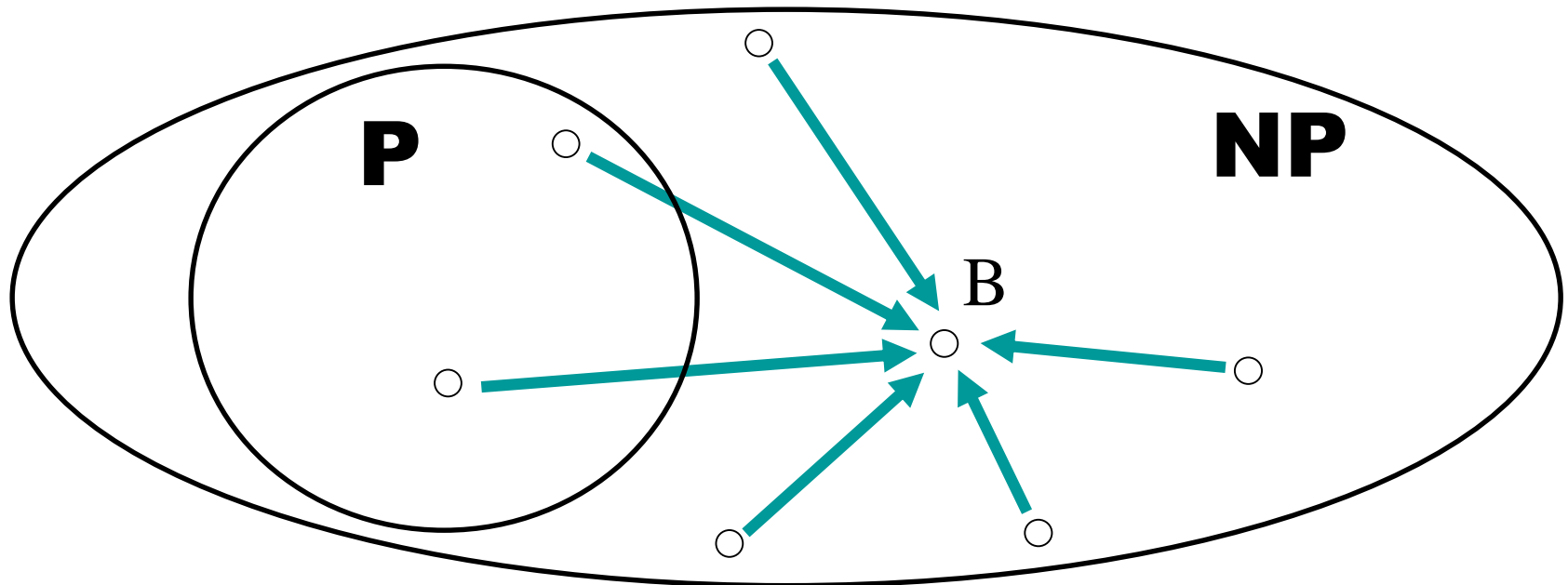
Run time. $O(k + \ell^2)$, i.e. polynomial in the input size.

- Basic reduction strategies.
 - Simple equivalence: $\text{INDEPENDENT-SET} \equiv_P \text{VERTEX-COVER}$.
 - Special case to general case: $\text{VERTEX-COVER} \leq_P \text{SET-COVER}$.
 - Encoding with gadgets: $3\text{-SAT} \leq_P \text{INDEPENDENT-SET}$.
- **Transitivity.** If $X \leq_P Y$ and $Y \leq_P Z$, then $X \leq_P Z$.
- Proof idea. Compose the two algorithms.
- Ex: $3\text{-SAT} \leq_P \text{INDEPENDENT-SET} \leq_P \text{VERTEX-COVER} \leq_P \text{SET-COVER}$.

Hardest problems in NP

A language B is **NP-complete** if

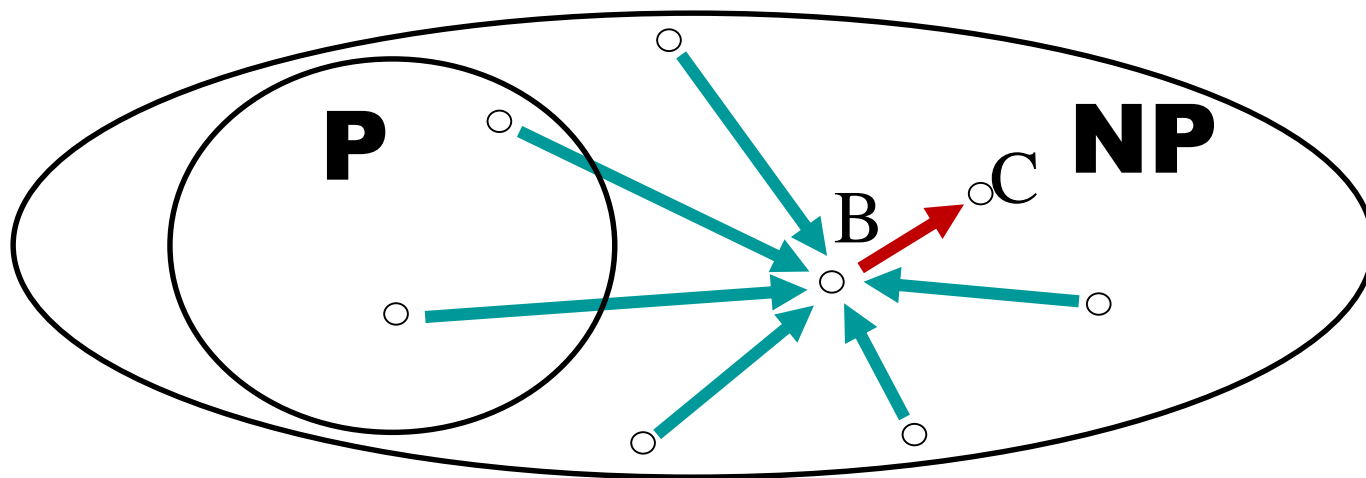
1. $B \in \text{NP}$
2. B is **NP-hard**, i.e., every language in NP is poly-time reducible to B.



Implication of poly-time reductions

Theorem. If

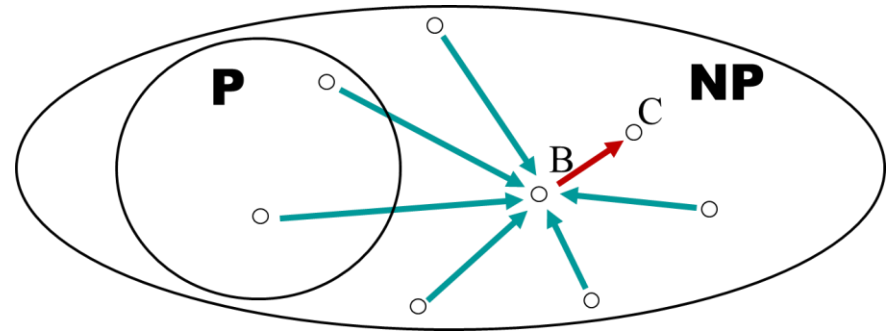
- B is **NP**-complete,
 - $C \in \mathbf{NP}$ and
 - $B \leq_p C$
- then C is **NP**-complete.



Implication of poly-time reductions

Theorem. If

- B is **NP**-complete,
 - $C \in \mathbf{NP}$ and
 - $B \leq_p C$
- then C is **NP**-complete.



Theorem. If B is **NP**-complete and $B \in \mathbf{P}$ then
 $\mathbf{P} = \mathbf{NP}$.

(So, if B is **NP**-complete and $\mathbf{P} \neq \mathbf{NP}$
then there is no poly-time algorithm for B.)

