## COMP 446 / 546 ALGORITHM DESIGN AND ANALYSIS

LECTURE 14 P-NP ALPTEKİN KÜPÇÜ

#### TRACTABLE PROBLEMS

- Our focus so far: Problems that can be solved efficiently
  - Matrix Multiplication
  - Is N a prime or composite
  - Max Flow Min Cut
  - Linear Programming
- Tractable: Problems that can be solved by algorithms which run in time O(n<sup>k</sup>) for some constant k > 0 where n is the input size.

### **INTRACTABLE PROBLEMS**

- Are all problems tractable? Far from it.
- No known polynomial-time algorithm for:
  - Traveling Salesman Problem: On a weighted graph G, find minimum-cost cycle that visits each vertex exactly once.
  - Factoring: On input N, find its prime factorization
- No known (any complexity) algorithm for:
  - Halting Problem: Given a program P (e.g., written in Java) and an input x, does P(x) terminate? UNDECIDABLE!!

#### **CLASSIFY PROBLEMS**

- Provably un-deciable: Halting
- Provably requires exponential-time:
  - Given a Turing machine, does it halt in at most n 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.

## DECISION, SEARCH, OPTIMIZATION PROBLEMS

#### Decision problem:

 Given a matrix A, vectors b and c, and a value k, is there a realvalued vector x, such that Ax ≤ b and cx ≥ k.

#### Search Problem:

 Given a matrix A, and vectors b and c, and a value k, find a realvalued vector x such that Ax ≤ b and cx ≥ k.

#### Optimization problem:

- Given a matrix A, and vectors b and c, find a real-valued vector x, such that Ax ≤ b that maximizes the value of cx.
- Optimization is harder than Search, which is harder than Decision
  - If one can solve Optimization, then one can easily solve Search
  - If one can solve Search, then one can easily solve Decision

## DECISION, SEARCH, OPTIMIZATION PROBLEMS

#### Decision problem:

 Given a weighted graph G=(V,E), a weight function w, and a bound k, is there a spanning tree of G of total weight less than k.

#### Search Problem:

 Given a weighted graph G=(V,E), a weight function w, and a bound k, find a spanning tree of G of total weight less than k.

#### Optimization problem:

 Given a weighted graph G=(V,E), and a weight function w, find a spanning tree of G of minimum possible total weight.

#### **DECISION PROBLEMS**

- Problems whose answer is either Yes or No
- Complexity theory usually deals with decision version of problems
  - We are interested in classifying how hard problems are.
  - If one can show decision version is hard, then we immediately know its search and optimization versions are also hard.
- A decision problem is a function D from the set all binary strings (all possible inputs represented in binary) to the set of {1,0}.
  - An input x to D is a yes-input if D(x)=1
  - An input x to D is a no-input if D(x)=0
  - We write x ∈ D if and only if D(x)=1

#### **COMPLEXITY CLASS P**

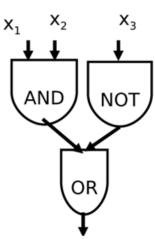
- Complexity Class P = { All decision problems D such that there exits a polynomial-time algorithm A such that A(x)=D(x) for all inputs x }
  - A is an algorithm that solves the decision problem D
  - We say that A decides D
- Examples
  - Existence of a Spanning Tree with cost at most k?
  - Feasibility of a Linear Program ?
  - Is N a prime number ?

# PROBLEMS NOT KNOWN TO BE IN P: TSP

- Decision version of Traveling Sales Person (TSP):
  - Given a complete weighted graph G and an integer k, is there a simple cycle that visits all nodes exactly once whose total weight is less than or equal to k.
- Search version: Find such a cycle
- Optimization version: Find minimum-weight simple cycle
- Best known algorithm: Solve decision version by solving the search version
  - Find cycle in time  $O(2^{\vee})$  for a graph with  $\vee$  vertices.
  - Given cycle, it is easy to verify that its weight is ≤ k
  - Given a cycle, one cannot verify its optimality in polynomial time
    - Search problem does not let us solve optimization problem easily

# PROBLEMS NOT KNOWN TO BE IN P: SAT & CIRCUIT-SAT

- Circuit Satisfiability (CIRCUIT-SAT): Given a Boolean circuit C made of gates and wires (an acyclic diagraph whose vertices can be AND/OR/NOT gates) with n inputs and one output, is there a way to assign 0-1 values to the inputs so that the output value is 1?
- General Satisfiability (SAT): Given a Boolean formula over n variables, is there an assignment of TRUE/FALSE to the variables that make the formula TRUE.
- For this lecture, we will not distinguish between SAT and cSAT.
- Best Known algorithm: Try out all 2<sup>n</sup> assignments.
  - EXPONENTIAL!!



## PROBLEMS NOT KNOWN TO BE IN P: FACTORING

- Decision version of Factoring: Given an n-bit integer N and bound B, is there a divisor d of N such that 1<d<B</li>
- Search Version: Given an n-bit integer N, find a divisor d of N such that 1<d<N</li>
- Best known algorithm:  $O(e^{K})$  where  $K = n^{1/3} \log_{n}^{2/3}$ 
  - Sub-exponential
  - Super-polynomial
- Interesting note: A quantum algorithm by Shor solves Factoring in polynomial-time on a quantum computer.

## PROBLEMS NOT KNOWN TO BE IN P: COMMON OBSERVATIONS

#### Solve search to solve decision:

- Finding the solution to the search version of these problems is hard
- But once you found it, the solution has polynomial size and can be verified to be correct in polynomial time

#### For a decision problem D

- For yes-inputs x of D, there is a short and easy to verify certificate that x ∈ D
- The certificate is simply the solution to the search variant of the problem
- Short means polynomial-size
- Easy to verify means polynomial-time verification

#### ONLY FOR YES-INPUTS !!!!!!!!!!!!

#### **COMPLEXITY CLASS NP**

- Complexity Class NP: All decision problems D for which there exist polynomial-time verification algorithm V and constant c > 0 such that
  - If D(x) = 1 (x is a yes-input of D) then ∃ certificate y s.t. |y| < |x|<sup>c</sup> and V(x,y) = ACCEPT
  - If D(x) = 0 (x is a no-input of D) then ∀ certificates y V(x,y) = REJECT
- NO-inputs cannot have certificates that cause the verification algorithm to accept.
- Verification algorithm never accepts NO-inputs regardless of the certificate provided.

#### **COMPLEXITY CLASS NP**

- Theoretical Non-Deterministic Turing Machine
  - Can "guess" the answer
  - Or, can be thought as running many copies of the program in parallel
  - Accepts an input x if there exists some sequence of operations that outputs "yes" on input x.
- NP = The class of problems that can be solved in polynomial-time on a Non-Deterministic Turing Machine.
  - Polynomial-time verification definition is equivalent!

### **REMARKS**

- For every yes-input for an NP problem, there exists at least one certificate, but possibly many
- P means decision problems that can be solved in polynomial time
- NP does not mean decision problems that are "not solvable in polynomial time"
- NP means "decision problems with polynomial-time verification algorithms" (non-deterministic polynomial time)

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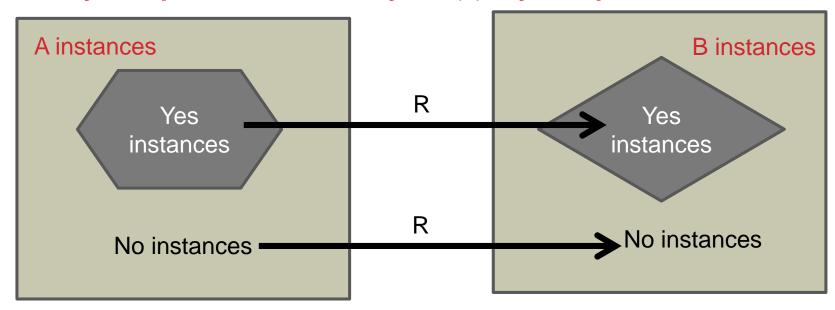
### **REMARKS**

- For every yes-input for an NP problem, there exists at least one certificate, but possibly many
- P means decision problems that can be solved in polynomial time
- NP does not mean decision problems that are "not solvable in polynomial time"
- NP means "decision problems with polynomial-time verification algorithms" (non-deterministic polynomial time)
- P ⊆ NP (every problem in P is also in NP)
  - Since every deterministic TM is also non-deterministic
- Not all decision problems are in NP.
  - Ex: Given circuit C decide if all assignments to the input variables make the output 0 (FALSE).
  - No short certificate is known that can be verified in polynomial-time

#### P vs. NP

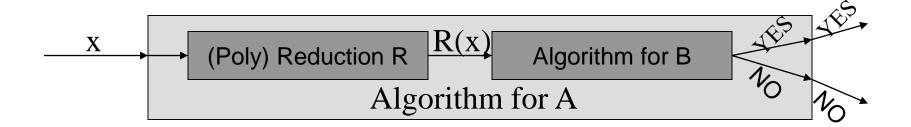
- P =? NP
  - Is it as easy to solve as to verify?
- Main problem: P and NP classify problems, not solutions
  - But, classification depends on solution algorithms
- No proof exists that shows a polynomial-time solution to an NP problem cannot exist.
  - But highly unlikely
  - The question has been open since 1970s
  - No such algorithm has been found
- How can we make progress on this problem?
  - Find the "hardest" problems in NP and just work on them
  - The NP-Complete problems

- Let A and B be decision problems
- A is polynomial-time reducible to B if there exists a polynomial-time algorithm R s.t.
  - R transforms input x to problem A into input R(x) to problem B so that x is a yes-input for A if and only if R(x) is yes-input for B



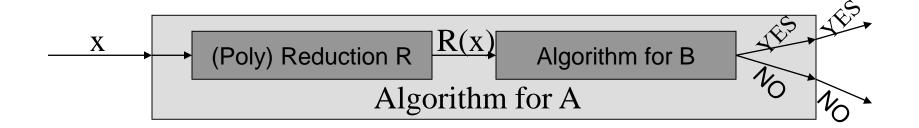
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Claim: If A is polynomial-time reducible to B and if ∃ polynomial-time algorithm for B then ∃ polynomial-time algorithm for A



#### Proof:

- Assume R takes p(|x|) time and the algorithm for B takes q(|y|) time, where p and q are polynomial functions.
- Further assume the output of R is r(|x|) bits where r is a polynomial.
- Then on input x the running time of A is p(|x|) + q(r(|x|)) = poly(|x|)



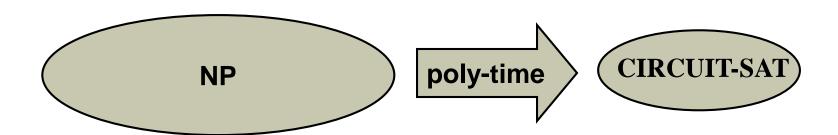
#### Main uses of reductions:

- Solving problem A using algorithm for B
- Showing B is harder than (at least as hard as) A
  - Because if we can solve B, we can easily solve A
  - But even if we can solve A, it may be hard to solve B
- Cryptographic security proofs

- Let B ≥ A (or A ≤ B) denote that A is polynomial-time reducible to B
  - We use ≥ because such a reduction means B is at least as hard as A.
- Reductions are transitive: If  $A \le B$  and  $B \le C$ , then  $A \le C$ .
  - An input x for A can be converted to x' for B, such that  $x \in A$  iff  $x' \in B$ .
  - Likewise, we can convert x' into x" for C such that  $x' \in B$  iff  $x'' \in C$ .
  - Hence, if  $x \in A$ , then  $x' \in B$  and  $x'' \in C$ .
  - Likewise, if  $x'' \in C$ , then  $x' \in B$  and  $x \in A$ .
  - Thus, A ≤ C since polynomials are closed under composition and addition. (required for sizes of certificates and verification times)

#### **NP-COMPLETENESS**

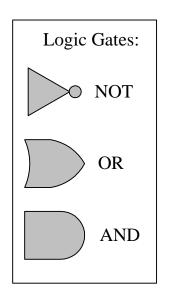
- A problem L is NP-hard if every problem in NP can be reduced to L in polynomial time.
- L is NP-complete if it is in NP and is NP-hard.
- Cook-Levin Theorem: SATISFIABILITY is NP-complete.
  - This includes SAT, CIRCUIT-SAT, 3-SAT

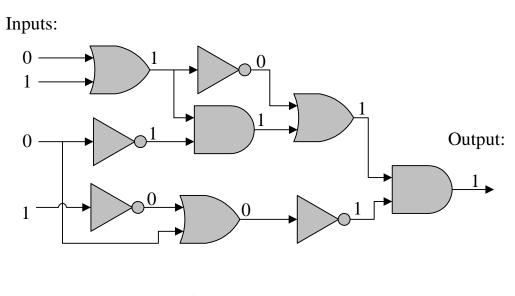


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#### SATISFIABILITY IS IN NP

- Consider a Boolean circuit composed of AND/OR/NOT gates.
- The CIRCUIT-SAT problem is to determine if there is an assignment of 0/1 to inputs so that the circuit outputs 1.
- Polynomial-time verification algorithm: Given a set of satisfying inputs (certificate), run each gate and check if the result is 1.

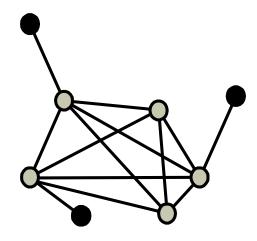




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## **CLIQUE**

- Input: An undirected graph G=(V,E) be and K > 0
- Problem: Is there a subset C of V with |C| ≥ K such that every pair of vertices in C has an edge between them?
  - Does the graph have a clique of size K?
  - Clique of size 3 are easy to detect
- Theorem: CLIQUE is NP-Complete.
- Proof:
  - First, obviously CLIQUE is in NP
    - Verification algorithm ?
  - For NP-hardness proof
    - Reduce 3-SAT to CLIQUE
    - Reduce inputs so that solving CLIQUE solves 3-SAT

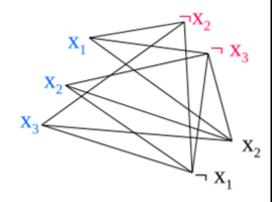


## 3SAT ≤ CLIQUE REDUCTION

- Given a 3-SAT input 3-CNF  $\phi = C_1, ..., C_m$  over  $x_1, ..., x_n$ , construct a CLIQUE input G = (V, E) and K s.t  $\phi$  satisfiable iff G has a clique of size  $\geq K$ 
  - Notation: a literal is either x<sub>i</sub> or ¬x<sub>i</sub>
- Reduction:

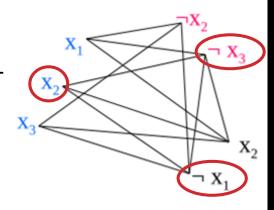
O(n+m)

- create a vertex for each literal t occuring in a clause
- create an edge between V<sub>t</sub> and V<sub>t</sub> unless
  - t and t' are in the same clause, or
  - t is the negation of t'
- set K=m (#clauses)
- Example:
  - Formula:  $(x_1 \lor x_2 \lor x_3) \land (\neg x_2 \lor \neg x_3) \land (\neg x_1 \lor x_2)$



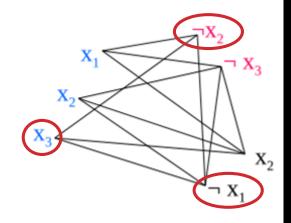
## 3SAT ≤ CLIQUE REDUCTION

- Claim: φ satisfiable iff G has a clique of size ≥ m
- φ satisfiable => G has a clique of size m
  - Take any assignment that satisfies φ
    - e.g., x<sub>1</sub>=F, x<sub>3</sub>=F, x<sub>2</sub>=T
  - Let the set of vertices C contain one literal for each clause that made the clause true
    - $X_2 \& \neg X_3 \& \neg X_1$
  - C is a clique of size m
    - All vertices in C are from different clauses
    - No two vertices' labels are negation of each other
      - Otherwise, not a valid assignment
        - No i with  $x_i = T$  and  $x_i = F$



## 3SAT ≤ CLIQUE REDUCTION

- Claim: φ satisfiable iff G has a clique of size ≥ m
- φ satisfiable <= G has a clique of size m</li>
  - Take any clique C of size m
    - All vertices in C are from different clauses
    - No two vertices' labels are negation of each other
  - Set all literals in the clique to evaluate to true in φ
    - e.g., x<sub>3</sub>=T, x<sub>2</sub>=F, x<sub>1</sub>=F
  - This is a legal assignment which satisfies φ

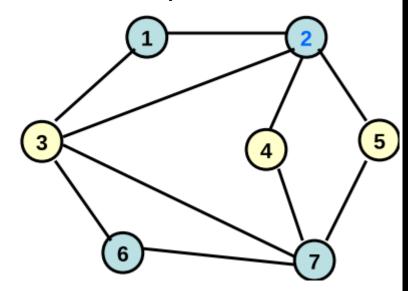


## 3SAT ≤ CLIQUE REDUCTION SUMMARY

- We constructed a reduction that maps:
  - YES-inputs of 3-SAT to YES-inputs of CLIQUE
  - NO-inputs of 3-SAT to NO-inputs of CLIQUE
- The reduction works in polynomial time
- Therefore, SAT ≤ CLIQUE and hence CLIQUE is NP-hard
- CLIQUE ∈ NP and CLIQUE is NP-hard
  - => CLIQUE is NP-complete

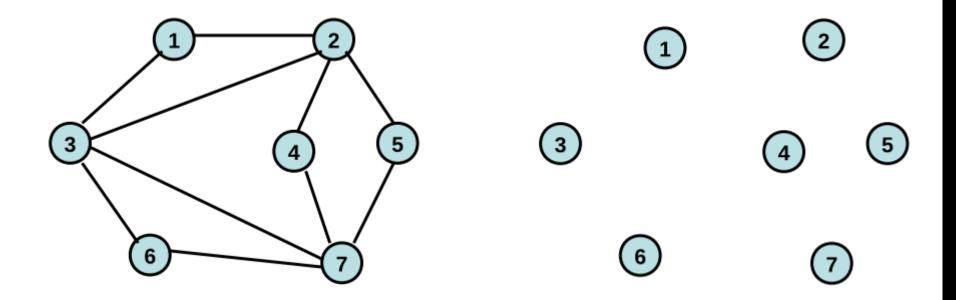
### **INDEPENDENT SET**

- Input: An undirected graph G=(V,E) and integer K > 0
- Problem: Is there a subset S of V of size K that is an independent set (that is there are no edges between vertices in S)
- Theorem: IS is NP-complete
- Proof:
  - First, obviously IS is in NP
    - Verification algorithm ?
  - For NP-hardness proof
    - Reduce CLIQUE to IS
    - Reduce inputs so that solving IS solves CLIQUE



#### **COMPLEMENT OF A GRAPH**

- G'=(V,E') is the complement of G=(V,E) if edge e ∈ E' iff e ∉ E
- Example: Construct the complement on board



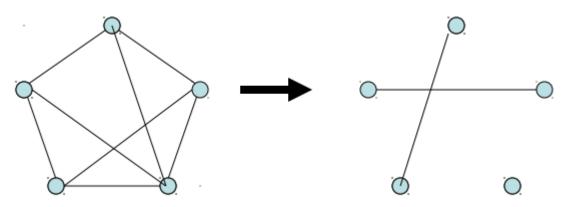
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### **CLIQUE ≤ INDEPENDENT SET**

 Claim: S is a clique in G iff S is an independent set in the complement of G

• Reduction: O(|V|<sup>2</sup>)

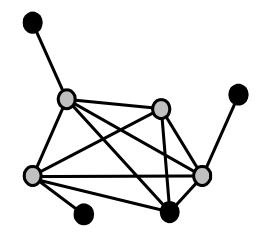
- Compute the complement of the graph G'.
- The complement graph has an independent set of size K iff the orginal graph had a clique of size K.
- (G,K) is reduced to (G',K)
- Duality again!



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### **VERTEX COVER**

- Input: An undirected graph G=(V,E) and an integer K > 0
- Problem: Is there a subset S of V that is a vertex cover of size at most K (every edge in E has at least one end point in S)
  - For all (a,b) ∈ E, a ∈ S or b ∈ S
- Theorem: Vertex Cover is NP-Complete
- Proof:
  - First, obviously VC is in NP
    - Verification algorithm ?
  - For NP-hardness proof
    - Reduce IS to VC
    - Reduce inputs so that solving VC solves IS



## INDEPENDENT SET ≤ VERTEX COVER

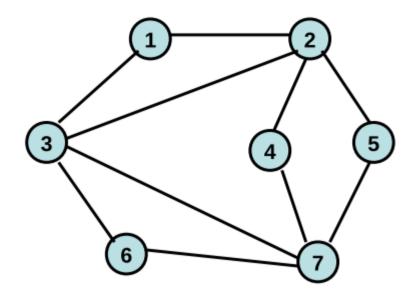
- Lemma: S is independent set in graph G iff V-S is a vertex cover in G
- Proof:
  - S independent set => V-S vertex cover
    - If S in independent set in G, there are no edges between vertices in S.
    - So all edges must have an end point in a vertex outside of S.
    - Thus, V-S is a vertex cover.
  - S independent set <= V-S vertex cover</li>
    - If V-S is a vertex cover in G, then all edges must have at least one end point in V-S.
    - So any pair of vertices out of V-S cannot have an edge between them otherwise that edge wouldn't be covered.
    - Thus, S is independent set.

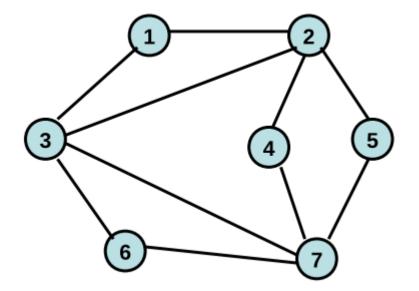
## INDEPENDENT SET ≤ VERTEX COVER

#### **EXERCISE**:

Find maximum independent set S

Show that V-S is a vertex cover

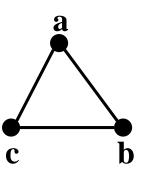




### 3-SAT ≤ VERTEX COVER

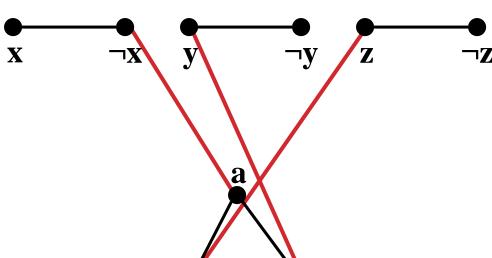
#### Reduce 3SAT to VERTEX-COVER:

- We have a Boolean formula in CNF with each clause having 3 literals.
- For each variable x, create a node for x and ¬x, and connect them with an edge:
- For each clause (a v b v c), create three nodes and connect them as a triangle.



### 3-SAT ≤ VERTEX COVER

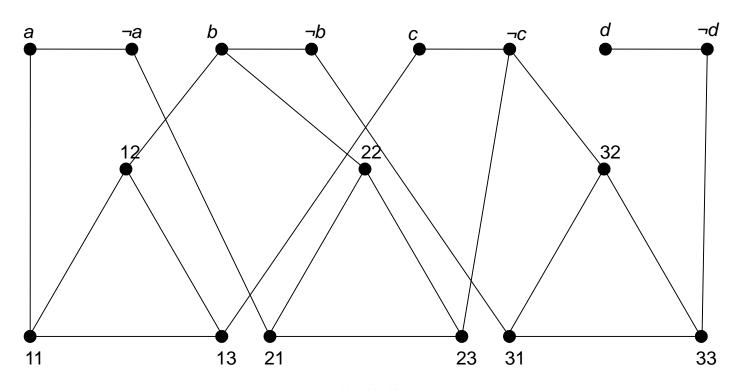
- Reduction (continued):
  - Connect each literal in a clause triangle to its copy in a variable pair.
    - e.g., a clause  $(\neg x \lor y \lor z)$



- Let n = # variables
- Let m = # clauses
- Set K=n+2m

#### 3-SAT ≤ VERTEX COVER

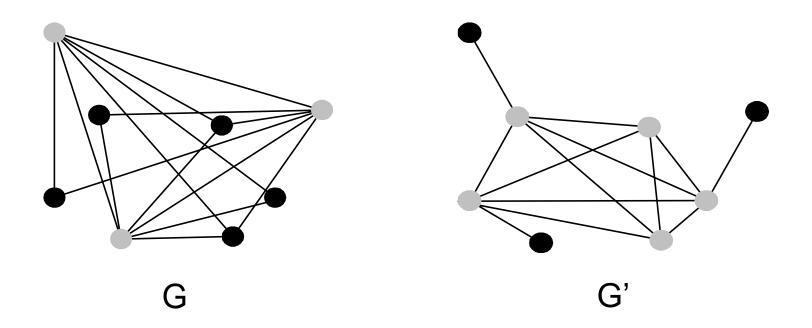
- Example: (a ∨ b ∨ c) ∧ (¬a ∨ b ∨ ¬c) ∧ (¬b ∨ ¬c ∨ ¬d)
- Graph has vertex cover of size K=4+6=10 iff formula is satisfiable.



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#### **VERTEX COVER ≤ CLIQUE**

- Reduce VERTEX-COVER to CLIQUE.
- A graph G has a vertex cover of size K if and only if its complement graph G' has a clique of size n-K.



### OTHER NP-COMPLETE PROBLEMS

- SET-COVER: Given a collection of m sets, are there K of these sets whose union is the same as the whole collection of m sets?
  - NP-complete by reduction from VERTEX-COVER
- SUBSET-SUM: Given a set of integers and a distinguished integer K, is there a subset of the integers that sums to K?
  - NP-complete by reduction from VERTEX-COVER

### OTHER NP-COMPLETE PROBLEMS

- 0-1 Knapsack: Given a collection of items with weights and benefits, is there a subset of weight at most W and benefit at least K?
  - NP-complete by reduction from SUBSET-SUM
- Hamiltonian-Cycle: Given an graph G, is there a cycle in G that visits each vertex exactly once?
  - NP-complete by reduction from VERTEX-COVER
- TSP: Given a complete weighted graph G, is there a simple cycle that visits each vertex and has total cost at most K?
  - NP-complete by reduction from Hamiltonian-Cycle.

I couldn't find a polynomial-time algorithm. I guess I'm not smart enough.



(cartoon inspired by [Garey-Johnson, 79])

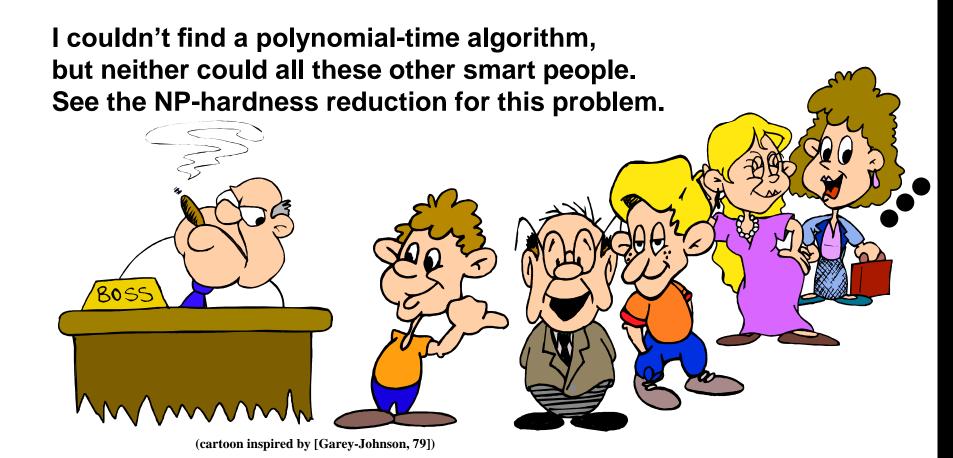
I couldn't find a polynomial-time algorithm, because no such algorithm exists!



Sometimes we can prove a strong lower bound... (but not usually)

(cartoon inspired by [Garey-Johnson, 79])

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- If you recognize your problem is NP-complete, do not waste your time trying to solve it in polynomial time.
- Options:
  - Use slow algorithm
  - Settle for an approximation of the solution
    - This applies to the search and optimization versions of the NPcomplete decision problem
  - Change your problem formulation so this updated version is in P.
    - Sometimes special cases are surprisingly easy.

#### REMARKS

- Special cases of a problem may be in P even if the general case is not known to be.
  - 2-SAT (P) vs. 3-SAT (NP-complete)
- Representation is important: Sometimes if inputs to a problem (or even just part of the input) is specified not in binary but say in unary, then it becomes much easier.
  - Clique when K is in unary can be solved in polynomial n<sup>K</sup> time.
  - Length of input K when represented in unary is K, instead of log K in binary.
- Fixed vs. Growing: Sometimes a portion of the input is really the same for all inputs (i.e., fixed), and that can make the problem tractable.
  - Integer programming with a fixed number of variables is in P
  - Clique with k=3 (i.e., deciding if there exists a triangle in the graph) is in P

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#### NP VS. NP-COMPLETENESS

- Graph Isomorphism: Given two graphs  $G_1=(V_1,E_1)$  and  $G_2=(V_2,E_2)$ , is there a mapping  $f: V_1 \to V_2$  such that  $(u,v) \in E_1$  iff  $(f(u), f(v)) \in E_2$
- Important Problem!
- Best known algorithm: O(2<sup>√V</sup>)
- Obviously in NP
  - Verification algorithm ?
- But probably not NP-complete
  - Definitely not known to be NP-complete
  - Similarly not known to be NP-complete: Factoring Problem

#### **BEYOND NP**

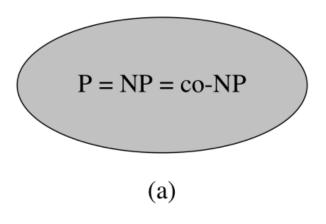
- Can you verify quickly that a 3-CNF formula is not satisfiable?
  - What would be a certificate?
- Can you verify quickly that a graph G does not have large clique?
  - What would be a certificate?
- Can you verify quickly that a pair of graphs are not isomorphic?
  - What would be a certificate?
- Best we know: exponential-size certificates for all these

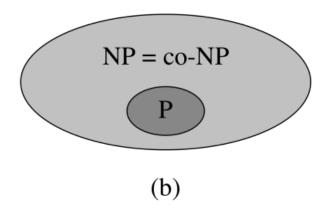
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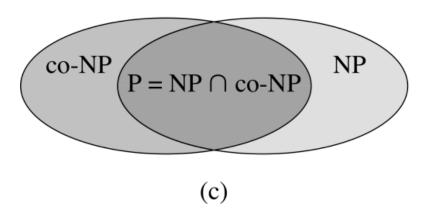
#### CO-NP

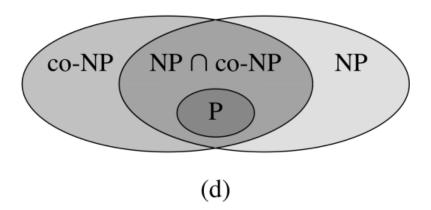
- NP: Decision problems that have a polynomial-size certificate that can be verified in polynomial-time for yes-instances.
  - Ex: SAT, HAM-CYCLE, COMPOSITES
- Complexity Class co-NP: Set containing complements of decision problems in NP. co-NP problems have a polynomial-size certificate that can be verified in polynomial-time for no-instances.
  - Ex: TAUTOLOGY, NO-HAM-CYCLE, PRIMES

# POSSIBLE OPTIONS (NONE PROVEN YET)









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