

## Two Notions

- Let  $R \subseteq \Sigma^* \times \Sigma^*$  be a binary relation on strings.
- $R$  is called **polynomially decidable** if

$$\{x; y : (x, y) \in R\}$$

is in P.

- $R$  is said to be **polynomially balanced** if  $(x, y) \in R$  implies  $|y| \leq |x|^k$  for some  $k \geq 1$ .

## An Alternative Characterization of NP

**Proposition 28 (Edmonds, 1965)** *Let  $L \subseteq \Sigma^*$  be a language. Then  $L \in NP$  if and only if there is a polynomially decidable and polynomially balanced relation  $R$  such that*

$$L = \{x : (x, y) \in R \text{ for some } y\}.$$

- Suppose such an  $R$  exists.
- $L$  can be decided by this NTM:
  - On input  $x$ , the NTM guesses a  $y$  of length  $\leq |x|^k$  and tests if  $(x, y) \in R$  in polynomial time.
  - It returns “yes” if the test is positive.

## The Proof (continued)

- Suppose that  $L \in \text{NP}$ .
- NTM  $N$  decides  $L$  in time  $|x|^k$ .
- Define  $R$  as follows:  $(x, y) \in R$  if and only if  $y$  is the encoding of an accepting computation of  $N$  on input  $x$ .
- Clearly  $R$  is polynomially bounded because  $N$  is polynomially bounded.
- $R$  is also polynomially decidable because it can be efficiently verified by simulation.
- Finally  $L = \{x : (x, y) \in R \text{ for some } y\}$  because  $N$  decides  $L$ .

## Comments

- Any “yes” instance  $x$  of an NP problem has at least one **succinct certificate** or **polynomial witness**  $y$  of its being a “yes” instance.
- “No” instances have none.
- Certificates are short and easy to verify.
  - An alleged satisfying truth assignment for SAT, an alleged Hamiltonian path for HAMILTONIAN PATH.
- Certificates may be hard to generate (otherwise, NP equals P), but verification must be easy.
- NP is the class of *easy-to-verify* problems.

## You Have an NP-Complete Problem (for Your Thesis)

- From Propositions 23 (p. 163) and Proposition 24 (p. 164), it is the least likely to be in P.
- Approximations.
- Special cases.
- Average performance.
- Randomized algorithms.
- Exponential-time algorithms that work well for small problems.
- “Heuristics” (and pray).

### 3SAT

- $k$ SAT, where  $k \in \mathbb{Z}^+$ , is the special case of SAT.
- The formula is in CNF and all clauses have *exactly*  $k$  literals (repetition of literals is allowed).
- For example,

$$(x_1 \vee x_2 \vee \neg x_3) \wedge (x_1 \vee x_1 \vee \neg x_2) \wedge (x_1 \vee \neg x_2 \vee \neg x_3).$$

## 3SAT Is NP-Complete

- Recall Cook's Theorem (p. 177) and the reduction of CIRCUIT SAT to SAT (p. 156).
- The resulting CNF has at most 3 literals for each clause.
  - This shows that 3SAT where each clause has at most 3 literals is NP-complete.
- Finally, duplicate one literal once or twice to make it a 3SAT formula.

## Another Variant of 3SAT

**Proposition 29** *3SAT is NP-complete for expressions in which each variable is restricted to appear at most three times, and each literal at most twice.*

- 3SAT here requires only that each clause has at most 3 literals.
- Consider a 3SAT expression in which  $x$  appears  $k$  times.



## The Proof (continued)

- Replace the first occurrence of  $x$  by  $x_1$ , the second by  $x_2$ , and so on, where  $x_1, x_2, \dots, x_k$  are  $k$  new variables.
- Add  $(\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_3) \wedge \dots \wedge (\neg x_k \vee x_1)$  to the expression  $(x_1 \Rightarrow x_2 \Rightarrow \dots \Rightarrow x_k \Rightarrow x_1)$ .
  - Each clause may have fewer than 3 clauses.
- The equivalent expression satisfies the condition for  $x$ .

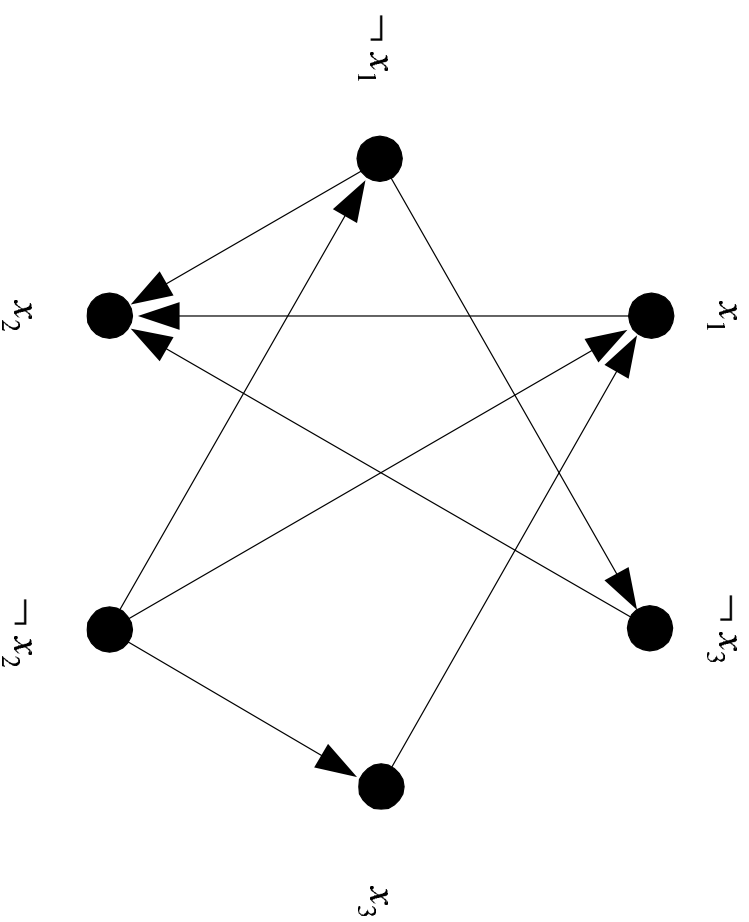
## 2SAT and Graphs

- Let  $\phi$  be an instance of 2SAT, in which each clause has exactly 2 literals.
- Define graph  $G(\phi)$  as follows:
  - The nodes are the variables and their negations.
  - Add edges  $(\neg\alpha, \beta)$  and  $(\neg\beta, \alpha)$  to  $G(\phi)$  if  $\alpha \vee \beta$  is a clause in  $\phi$ .
    - \* For example, if  $x \vee \neg y \in \phi$ , add  $(\neg x, \neg y)$  and  $(y, x)$ .
    - \* *Two* edges are added for each clause.
  - Think of the edges as  $\neg\alpha \Rightarrow \beta$  and  $\neg\beta \Rightarrow \alpha$ .
  - $b$  is reachable from  $a$  iff  $\neg a$  is reachable from  $\neg b$ .
  - Paths in  $G(\phi)$  are valid implications.

## Illustration

Digraph for

$$(x_1 \vee x_2) \wedge (x_1 \vee \neg x_3) \wedge (\neg x_1 \vee x_2) \wedge (x_2 \vee x_3).$$



## Properties of $G(\phi)$

**Theorem 30**  $\phi$  is unsatisfiable if and only if there is a variable  $x$  such that there are paths from  $x$  to  $\neg x$  and from  $\neg x$  to  $x$  in  $G(\phi)$ .

- Suppose that such paths exist, but  $\phi$  can be satisfied by a truth assignment  $T$ .
  - Without loss of generality, assume  $T(x) = \text{true}$ .
  - As there is a path from  $x$  to  $\neg x$  and  $T(\neg x) = \text{false}$ , there must be an edge  $(\alpha, \beta)$  on this path such that  $T(\alpha) = \text{true}$  and  $T(\beta) = \text{false}$ .
  - Hence  $(\neg\alpha \vee \beta)$  is a clause of  $\phi$ .
  - But this clause is *not* satisfied by  $T$ , a contradiction.

## The Proof (continued)

- Suppose there is no variable with such paths in  $G(\phi)$ .
- We shall construct a satisfying truth assignment.
- It is enough that no edges go from true to false.
- Pick any node  $\alpha$  which has not had a truth value and there is no path from it to  $\neg\alpha$  (always doable by assumption, why?).
- Assign nodes reachable from  $\alpha$  true and their negations false.
  - The negations are those nodes that can reach  $\neg\alpha$ .

## The Proof (continued)

- The above steps are well-defined.
  - If  $\alpha$  could reach both  $\beta$  and  $\neg\beta$ , then there would be a path from  $\neg\beta$  to  $\neg\alpha$ , hence a path from  $\alpha$  to  $\neg\alpha$ !
  - If there were a path from  $\alpha$  to a node  $y$  already assigned false, then  $\neg y$  can reach  $\neg\alpha$  and  $\alpha$  has been assigned false before!
- We keep picking such  $\alpha$ 's until we run out of them.
- Every node must have had a truth value.
  - If  $\alpha$  does not, it must be because there is a path from it to  $\neg\alpha$ , but then the algorithm could have picked  $\neg\alpha$ !
- The assignments make sure a false never follows a true.

## 2SAT Is in NL $\subseteq$ P

- By Corollary 21 on p. 145,  $\text{coNL}$  equals NL.
- We need to show only that recognizing unsatisfiable expressions is in NL.
- In nondeterministic logarithmic space, we can test the conditions of Theorem 30 by guessing a variable  $x$  and testing if  $\neg x$  is reachable from  $x$  and if  $\neg x$  can reach  $x$ .
  - See the algorithm for REACHABILITY (p. 70).

## Generalized 2SAT: MAX2SAT

- Consider a CNF in which all clauses have two literals.
- Let  $K \in \mathbb{N}$ .
- MAX2SAT is the problem of whether there is a truth assignment that satisfies at least  $K$  of the clauses.
- MAX2SAT becomes 2SAT when  $K$  equals the number of clauses.
- MAX2SAT is an optimization problem.
- MAX2SAT is in NP: Guess a truth assignment and verify the count.



## MAX2SAT Is NP-Complete<sup>a</sup>

- Consider the following 10 clauses:

$$(x) \wedge (y) \wedge (z) \wedge (w)$$

$$(\neg x \vee \neg y) \wedge (\neg y \vee \neg z) \wedge (\neg z \vee \neg x)$$

$$(x \vee \neg w) \wedge (y \vee \neg w) \wedge (z \vee \neg w)$$

- Let the 2SAT formula  $r(x, y, z, w)$  represent the conjunction of these clauses.
- How many clauses can we satisfy?
- The clauses are symmetric with respect to  $x$ ,  $y$ , and  $z$ .

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<sup>a</sup>Garey, Johnson, Stockmeyer, 1976.

## The Proof (continued)

**All of  $x, y, z$  are true:** By setting  $w$  to true, we *can* satisfy  $4 + 0 + 3 = 7$  clauses.

**Two of  $x, y, z$  are true:** By setting  $w$  to true, we *can* satisfy  $3 + 2 + 2 = 7$  clauses; by setting  $w$  to false, we *can* satisfy  $2 + 2 + 3 = 7$  clauses.

**One of  $x, y, z$  is true:** By setting  $w$  to false, we *can* satisfy  $1 + 3 + 3 = 7$  clauses, whereas by setting  $w$  to true, we *can* satisfy only  $2 + 3 + 1 = 6$  clauses.

**None of  $x, y, z$  is true:** By setting  $w$  to false, we *can* satisfy  $0 + 3 + 3 = 6$  clauses, whereas by setting  $w$  to true, we *can* satisfy only  $1 + 3 + 0 = 4$  clauses.

## The Proof (continued)

- Any truth assignment that satisfies  $x \vee y \vee z$  can be extended to satisfy 7 of the 10 clauses and no more.
- The remaining truth assignment can be extended to satisfy only 6 of them.
- The reduction from 3SAT  $\phi$  to MAX2SAT  $R(\phi)$ :
  - For each clause  $C_i = (\alpha \vee \beta \vee \gamma)$  of  $\phi$ , add **group**  $r(\alpha, \beta, \gamma, w_i)$  to  $R(\phi)$ .
  - If  $\phi$  has  $m$  clauses, then  $R(\phi)$  has  $10m$  groups.
- Set  $K = 7m$ .

## The Proof (continued)

- We now show that  $K$  clauses of  $R(\phi)$  can be satisfied if and only if  $\phi$  is satisfiable.
- Suppose  $7m$  clauses of  $R(\phi)$  can be satisfied.
  - $7$  clauses must be satisfied in each group because each group can only have at most  $7$  clauses satisfied.
  - But all clauses *in*  $\phi$  must be satisfied.
- Suppose all clauses of  $\phi$  are satisfied.
  - Each group can set its  $w_i$  appropriately to have  $7$  clauses satisfied.

## NAESAT

- The NAESAT (for “not-all-equal” SAT) is like 3SAT.
- But we require additionally that there be a satisfying truth assignment under which no clauses have the three literals equal in truth value.
  - Each clause must have one literal assigned true and one literal assigned false.

## NAESAT Is NP-Complete<sup>a</sup>

- Recall the reduction of CIRCUIT SAT to SAT on p. 156.
- It produced a CNF  $\phi$  in which each clause has at most 3 literals.
- Add the same variable  $z$  to all clauses with fewer than 3 literals to make it a 3SAT formula.
- We will argue that the new formula  $\phi(z)$  is NAE-satisfiable if and only if the original circuit is satisfiable.

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<sup>a</sup>Karp, 1972.

## The Proof (continued)

- Suppose  $T$  NAE-satisfies  $\phi(z)$ .
  - $\bar{T}$  also NAE-satisfies  $\phi(z)$ .
  - Under either  $T$  or  $\bar{T}$ , variable  $z$  takes the value false.
  - This truth assignment must satisfy all clauses of  $\phi$ .
  - So it satisfies the original circuit.

## The Proof (continued)

- Suppose there is a truth assignment that satisfies the circuit.
  - Then there is a truth assignment  $T$  that satisfies every clause of  $\phi$ .
  - Extend  $T$  by adding  $T(z) = \text{false}$  to obtain  $T'$ .
  - $T'$  satisfies  $\phi(z)$ .
  - So in no clauses are all three literals false under  $T'$ .
  - Under  $T'$ , in no clauses are all three literals true.
- \* Review the construction on p. 157 and p. 158.



## Undirected Graphs

- An **undirected graph**  $G = (V, E)$  has a finite set of nodes,  $V$ , and a set of *undirected edges*,  $E$ .
- It is like a graph except that the edges have no directions and there are no self-loops.
- We use  $[i, j]$  to denote the fact that there is an edge between node  $i$  and node  $j$ .

## Independent Sets

- Let  $G = (V, E)$  be an undirected graph.
- $I \subseteq V$ .
- $I$  is **independent** if whenever  $i, j \in I$ , there is no edge between  $i$  and  $j$ .
- The INDEPENDENT SET problem is this: Given an undirected graph and a goal  $K$ , is there an independent set of size  $K$ ?
  - Many applications.

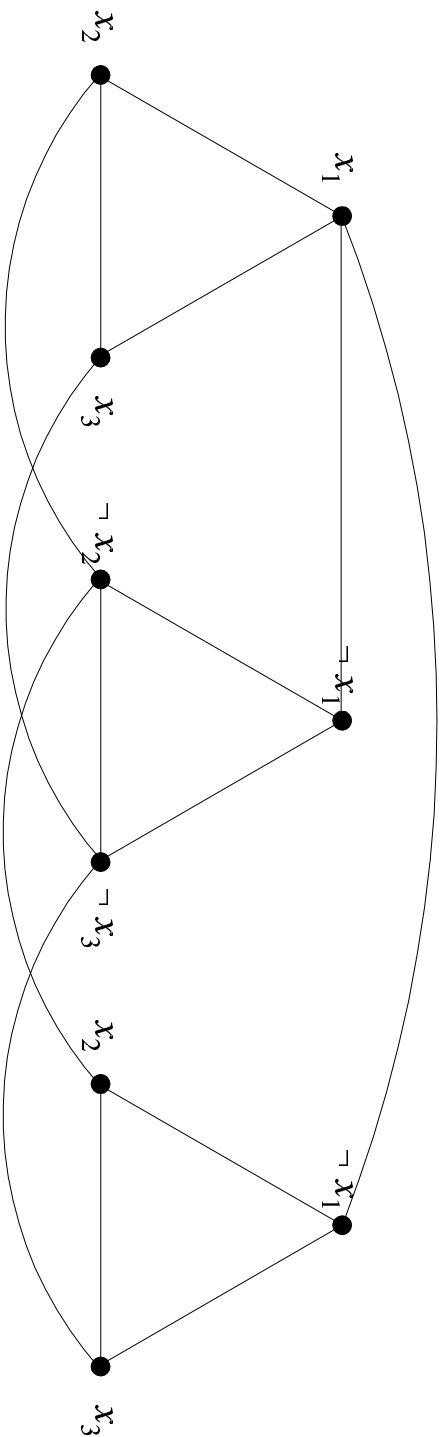
## INDEPENDENT SET Is NP-Complete

- This problem is in NP: Guess a set of nodes and verify that it is independent and meets the count.
- If a graph contains a triangle, any independent set can contain at most one node of the triangle.
- We consider graphs whose nodes can be partitioned in  $m$  disjoint triangles.
  - If the subproblem is hard, the original problem is at least as hard.

## Reduction from 3SAT to INDEPENDENT SET

- Let  $\phi$  be an instance of 3SAT with  $m$  clauses.
- We will construct graph  $G$  (with constraints as said) with  $K = m$  such that  $\phi$  is satisfiable if and only if  $G$  has an independent set of size  $K$ .
- There is a triangle for each clause with the literals as the nodes.
- Add additional edges between  $x$  and  $\neg x$  for every variable  $x$ .

## A Sample Construction



$$(x_1 \vee x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee x_3).$$

## The Proof (continued)

- Suppose  $G$  has an independent set  $I$  of size  $K = m$ .
  - An independent set can contain at most  $m$  nodes, one from each triangle.
  - An independent set of size  $m$  exists if and only if it contains exactly one node from each triangle.
  - Truth assignment  $T$  assigns true to those literals in  $I$ .
  - $T$  is consistent because contradictory literals are connected by an edge, hence not both in  $I$ .
  - $T$  satisfies  $\phi$  because it has a node from every triangle, thus satisfying every clause.

## The Proof (continued)

- Suppose a satisfying truth assignment  $T$  exists for  $\phi$ .
  - Collect one node from each triangle whose literal is true under  $T$ .
  - This set of  $m$  nodes must be independent by construction.

**Corollary 31** 4-DEGREE<sup>a</sup> INDEPENDENT SET is *NP-complete*.

**Theorem 32** INDEPENDENT SET is *NP-complete for planar graphs*.

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<sup>a</sup>The degrees in the graph are at most 4 if we start with NAESAT.

## CLIQUE and NODE COVER

- We are given an undirected graph  $G$  and a goal  $K$ .
- CLIQUE asks if there is a set of  $K$  nodes that form a **clique**, which have all possible edges between them.
- NODE COVER asks if there is a set  $C$  with  $K$  or fewer nodes such that each edge of  $G$  has at least one of its endpoints in  $C$ .



## Both CLIQUE and NODE COVER Are NP-Complete

**Corollary 33** CLIQUE is NP-complete.

- Let  $\bar{G}$  be the complement of  $G$ , where  $[x, y] \in \bar{G}$  if and only if  $[x, y] \notin G$ .
- Then  $I$  is a clique in  $G$  if and only if  $I$  is an independent set in  $\bar{G}$ .

**Corollary 34** NODE COVER is NP-complete.

- $I$  is an independent set of  $G = (V, E)$  if and only if  $V - I$  is a node cover of  $G$ .

## MIN CUT and MAX CUT

- A **cut** in an undirected graph  $G = (V, E)$  is a partition of the nodes into two nonempty sets  $S$  and  $V - S$ .
- The size of a cut  $(S, V - S)$  is the number of edges between  $S$  and  $V - S$ .
- MIN CUT is in P.
- MAX CUT asks if there is a cut of size at least  $K$ .