

Another Variant of 3SAT

Proposition 42 *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.)*

The Proof (continued)

- Consider a general 3SAT expression in which x appears k times.
- Replace the first occurrence of x by x_1 , the second by x_2 , and so on.
 - x_1, x_2, \dots, x_k are k new variables.

The Proof (concluded)

- Add $(\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_3) \wedge \cdots \wedge (\neg x_k \vee x_1)$ to the expression.
 - It is logically equivalent to

$$x_1 \Rightarrow x_2 \Rightarrow \cdots \Rightarrow x_k \Rightarrow x_1.$$

- So it is necessary that x_1, x_2, \dots, x_k assume an identical truth value to satisfy the whole expression.
- Note that each clause $\neg x_i \vee x_j$ above has only 2 literals.
- The resulting equivalent expression satisfies the conditions for x .

An Example

- Suppose we are given the following 3SAT expression

$$\cdots (\neg x \vee w \vee g) \wedge \cdots \wedge (x \vee y \vee z) \cdots .$$

- The transformed expression is

$$\cdots (\boxed{\neg x_1} \vee w \vee g) \wedge \cdots \wedge (\boxed{x_2} \vee y \vee z) \cdots (\boxed{\neg x_1} \vee \boxed{x_2}) \wedge (\boxed{\neg x_2} \vee \boxed{x_1}).$$

- Variable x_1 appears 3 times.
- Literal x_1 appears once.
- Literal $\neg x_1$ appears 2 times.

2SAT Is in $NL \subseteq P$

- Let ϕ be an instance of 2SAT: Each clause has 2 literals.
- NL is a subset of P.^a
- Because $coNL = NL$,^b we need to show only that recognizing *unsatisfiable* 2SAT expressions is in NL.
- See the textbook for the complete proof.

^aRecall p. 255.

^bRecall p. 269.

Generalized 2SAT: MAX2SAT

- Consider a 2SAT formula.
- Let $K \in \mathbb{N}$.
- MAX2SAT asks whether there is a truth assignment that satisfies at least K of the clauses.
 - MAX2SAT becomes 2SAT when K equals the number of clauses.

Generalized 2SAT: MAX2SAT (concluded)

- MAX2SAT can be used to solve the related optimization version.
 - With binary search, one can nail the maximum number of satisfiable clauses of 2SAT formulas.
- MAX2SAT \in NP: Guess a truth assignment and verify the count.
- We now reduce 3SAT to MAX2SAT.

MAX2SAT Is NP-Complete^a

- 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.
- The clauses are symmetric with respect to x , y , and z .
- How many clauses can we satisfy?

^aGarey, Johnson, & Stockmeyer (1976).

The Proof (continued)

All of x, y, z are true: By setting w to true, we satisfy $4 + 0 + 3 = 7$ clauses, whereas by setting w to false, we satisfy only $3 + 0 + 3 = 6$ clauses.

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

The Proof (continued)

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

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

The Proof (continued)

- A truth assignment that satisfies $x \vee y \vee z$ can be *extended* to satisfy 7 of the 10 clauses of $r(x, y, z, w)$, *and no more*.
- A truth assignment that does *not* satisfy $x \vee y \vee z$ can be extended to satisfy only 6 of them, *and no more*.
- The reduction from 3SAT ϕ to 2SAT $R(\phi)$:
 - For each clause $C_i = (\alpha \vee \beta \vee \gamma)$ of ϕ , add $r(\alpha, \beta, \gamma, w_i)$ to $R(\phi)$.
- If ϕ has m clauses, then $R(\phi)$ has $10m$ clauses.

The Proof (continued)

- Finally, set $K = 7m$.
- So the reduction transforms ϕ to $(R(\phi), 7m)$.
- We now show that K clauses of $R(\phi)$ can be satisfied if and only if ϕ is satisfiable.

The Proof (continued)

- Suppose $K = 7m$ clauses of $R(\phi)$ can be satisfied.
 - 7 clauses of each $r(\alpha, \beta, \gamma, w_i)$ must be satisfied because it can have at most 7 clauses satisfied.^a
 - Hence each clause $C_i = (\alpha \vee \beta \vee \gamma)$ of ϕ is satisfied by the same truth assignment.
 - So ϕ is satisfied.

^aIf 70% of the world population are male and if at most 70% of each country's population are male, then each country must have exactly 70% male population.

The Proof (concluded)

- Suppose ϕ is satisfiable.
 - Let T satisfy all clauses of ϕ .
 - Each $r(\alpha, \beta, \gamma, w_i)$ can set its w_i appropriately to have 7 clauses satisfied.
 - So $K = 7m$ clauses are satisfied.

NAESAT

- The NAESAT (for “not-all-equal” SAT) is like 3SAT.
- But there must be a satisfying truth assignment under which no clauses have all three literals equal in truth value.
 - Equivalently, there is a truth assignment such that each clause has a literal assigned true and a literal assigned false.
 - Equivalently, there is a *satisfying* truth assignment under which each clause has a literal assigned false.

NAESAT (concluded)

- Take

$$\begin{aligned}\phi &= (\neg x_1 \vee \neg x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee \neg x_3) \\ &\quad \wedge (x_1 \vee x_2 \vee x_3)\end{aligned}$$

as an example.

- Then $\{x_1 = \text{true}, x_2 = \text{false}, x_3 = \text{false}\}$
NAE-satisfies ϕ because

$$\begin{aligned}&(\boxed{\text{false}} \vee \text{true} \vee \text{true}) \wedge (\boxed{\text{false}} \vee \boxed{\text{false}} \vee \text{true}) \\ &\wedge (\text{true} \vee \boxed{\text{false}} \vee \boxed{\text{false}}).\end{aligned}$$

NAESAT Is NP-Complete^a

- Recall the reduction of CIRCUIT SAT to SAT on p. 293ff.
- It produced a CNF ϕ in which each clause had 1, 2, or 3 literals.
- Add the same variable z to all clauses with fewer than 3 literals to make it a 3SAT formula.
- Goal: The new formula $\phi(z)$ is NAE-satisfiable if and only if the original circuit is satisfiable.

^aSchaefer (1978).

The Proof (continued)

- The following simple observation will be useful.
- Suppose T NAE-satisfies a boolean formula ϕ .
- Let \bar{T} take the opposite truth value of T on every variable.
- Then \bar{T} also NAE-satisfies ϕ .^a

^aHesse's *Siddhartha* (1922), "The opposite of every truth is just as true!"

The Proof (continued)

- Suppose T NAE-satisfies $\phi(z)$.
 - \bar{T} also NAE-satisfies $\phi(z)$.
 - Under T or \bar{T} , variable z takes the value false.
 - *This* truth assignment \mathcal{T} must satisfy all the clauses of ϕ .
 - * Because z is not the reason that makes $\phi(z)$ true under \mathcal{T} anyway.
 - So $\mathcal{T} \models \phi$.
 - And the original circuit is satisfiable.

The Proof (concluded)

- Suppose there is a truth assignment that satisfies the circuit.
 - Then there is a truth assignment T that satisfies every clause of ϕ .
 - Extend T by adding $T(z) = \mathbf{false}$ to obtain T' .
 - T' satisfies $\phi(z)$.
 - Clearly, in no clauses are all three literals false under T' .
 - In no clauses are all three literals true under T' .
 - * Need to go over the detailed construction on pp. 294–296.

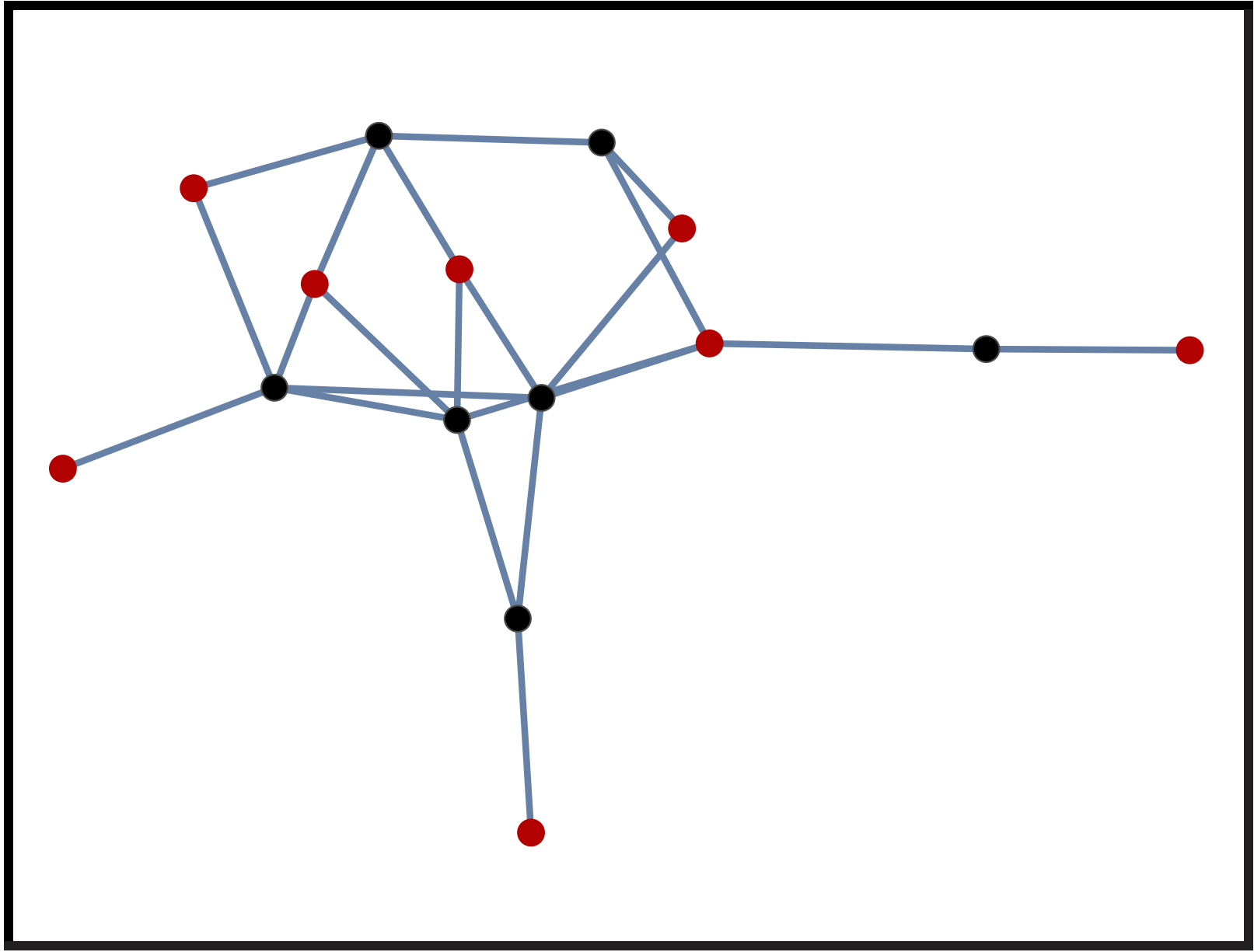
Undirected Graphs

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

^aAn equally good notation is $\{i, j\}$.

Independent Sets

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



INDEPENDENT SET Is NP-Complete

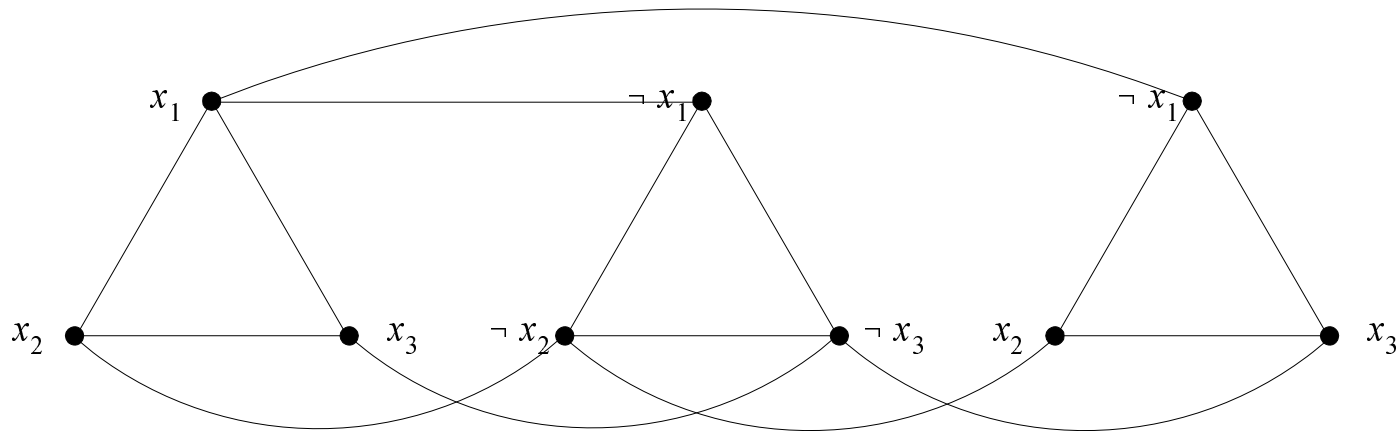
- This problem is in NP: Guess a set of nodes and verify that it is independent and meets the count.
- We will reduce 3SAT to INDEPENDENT SET.
- Note: If a graph contains a triangle, any independent set can contain at most one node of the triangle.
- The reduction will output graphs whose nodes can be partitioned into disjoint triangles, one for each clause.^a

^aRecall that a reduction does not have to be an onto function.

The Proof (continued)

- Let ϕ be a 3SAT formula with m clauses.
- We will construct graph G with $K = m$.
- Furthermore, ϕ is satisfiable if and only if G has an independent set of size K .
- Here is the reduction:
 - There is a triangle for each clause with the literals as the nodes' labels.
 - Add edges between x and $\neg x$ for every variable x .

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



Identical literals that appear in the same clause or different clauses yield *distinct* nodes.

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.
 - So I 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 both cannot be in I .
 - T satisfies ϕ because it has a node from every triangle, thus satisfying every clause.^a

^aThe variables without a truth value can be assigned arbitrarily. Contributed by Mr. Chun-Yuan Chen (R99922119) on November 2, 2010.

The Proof (concluded)

- Suppose ϕ is satisfiable.
 - Let truth assignment T satisfy ϕ .
 - Collect one node from each triangle whose literal is true under T .
 - The choice is arbitrary if there is more than one true literal.
 - This set of m nodes must be independent by construction.
 - * Because both literals x and $\neg x$ cannot be assigned true.

Other INDEPENDENT SET-Related NP-Complete Problems

Corollary 43 INDEPENDENT SET *is NP-complete for 4-degree graphs.*

Theorem 44 INDEPENDENT SET *is NP-complete for planar graphs.*

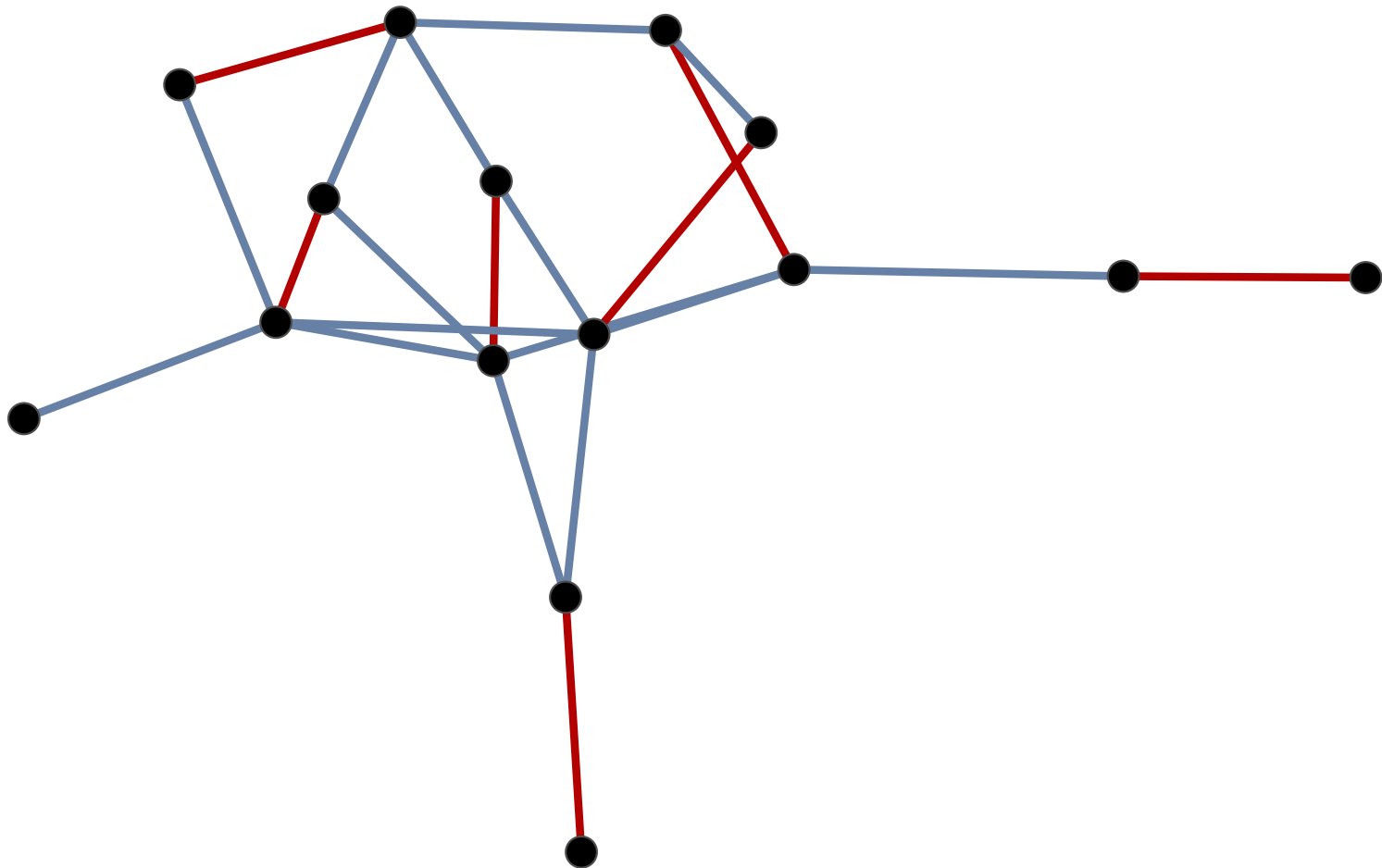
Theorem 45 (Garey & Johnson, 1977) INDEPENDENT SET *is NP-complete for 3-degree planar graphs.*

Is INDEPENDENT EDGE SET Also NP-Complete?

- INDEPENDENT EDGE SET: Given an undirected graph and a goal K , is there an independent *edge* set of size K ?
- This problem is equivalent to maximum matching!
- Maximum matching can be solved in polynomial time.^a

^aEdmonds (1965); Micali & V. Vazirani (1980).

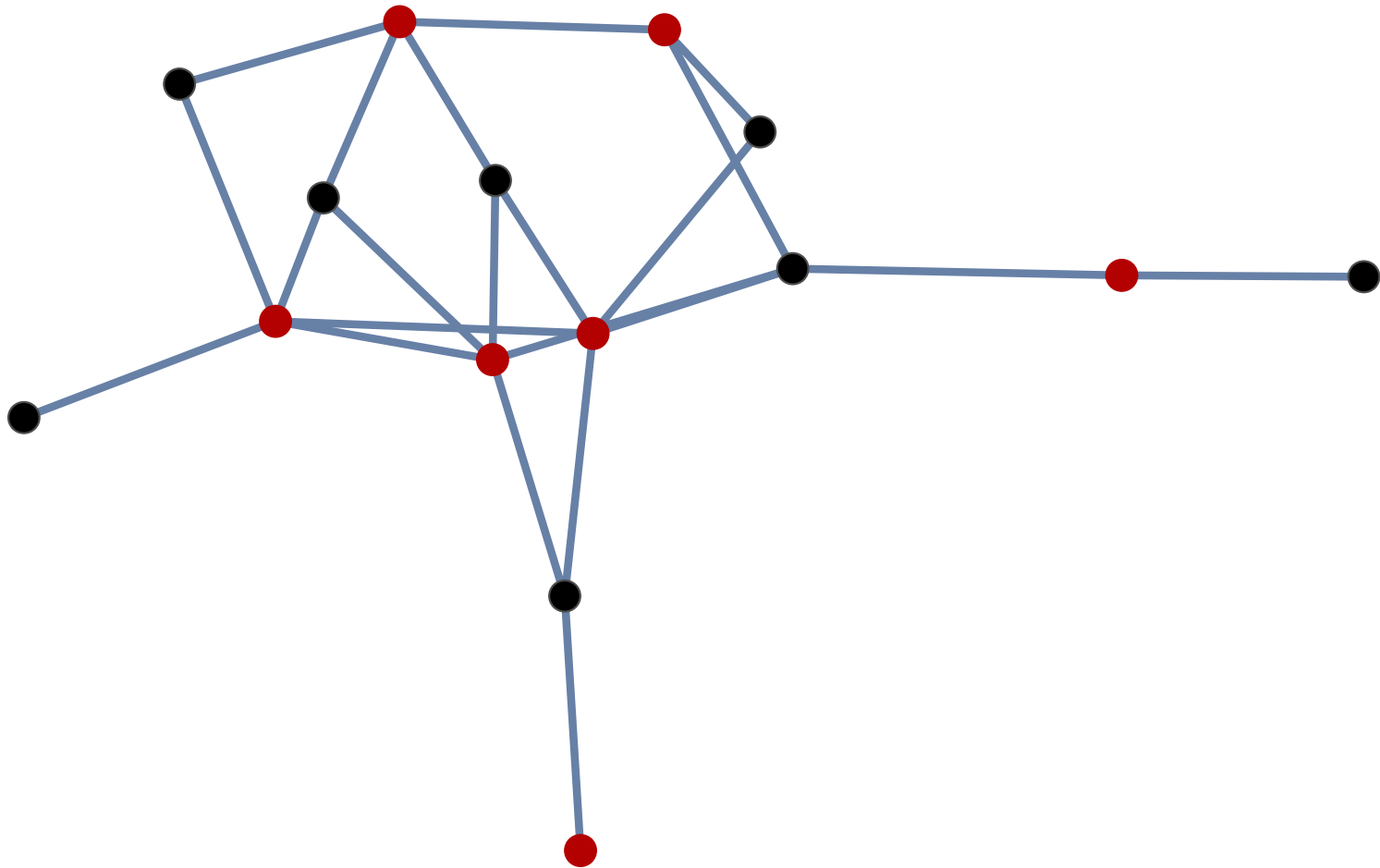
A Maximum Matching



NODE COVER

- We are given an undirected graph G and a goal K .
- NODE COVER: Is there a set C with K or fewer nodes such that each edge of G has at least one of its endpoints (i.e., incident nodes) in C ?
- Many applications.

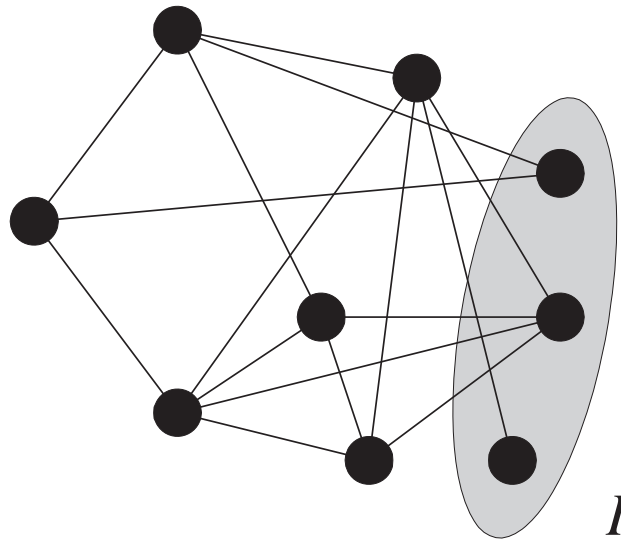
NODE COVER (concluded)



NODE COVER Is NP-Complete

Corollary 46 (Karp, 1972) 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 .^a



^aFinish the reduction!

Richard Karp^a (1935–)



^aTuring Award (1985).

Remarks^a

- Are INDEPENDENT SET and NODE COVER in P if K is a constant?
 - Yes, because one can do an exhaustive search on all the possible node covers or independent sets (both $\binom{n}{K} = O(n^K)$ of them, a polynomial).^b
- Are INDEPENDENT SET and NODE COVER NP-complete if K is a linear function of n ?
 - INDEPENDENT SET with $K = n/3$ and NODE COVER with $K = 2n/3$ remain NP-complete by our reductions.

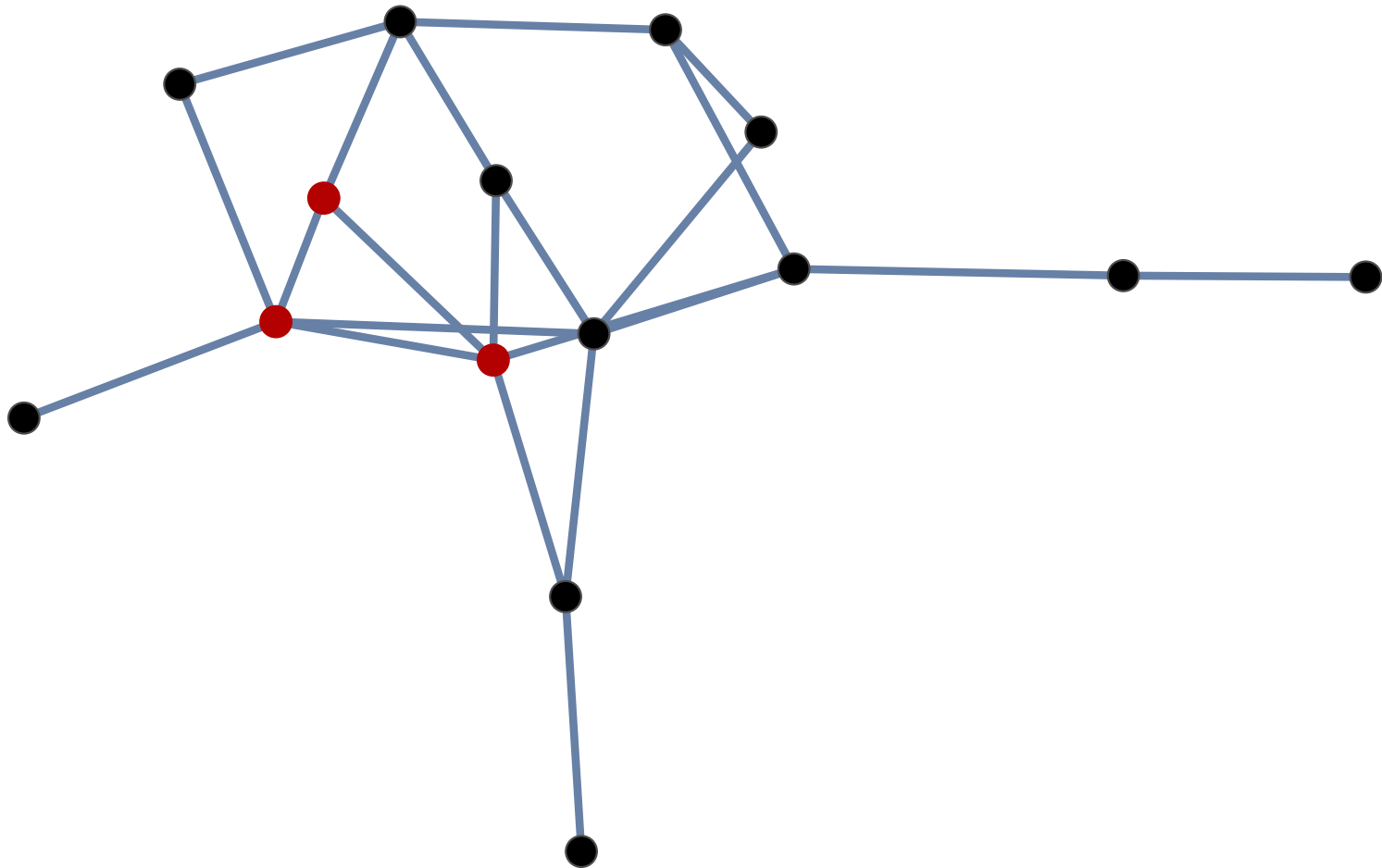
^aContributed by Mr. Ching-Hua Yu (D00921025) on October 30, 2012.

^b $n = |V|$.

CLIQUE

- We are given an undirected graph G and a goal K .
- CLIQUE asks if there is a set C with K nodes such that there is an edge between any two nodes $i, j \in C$.
- Many applications.

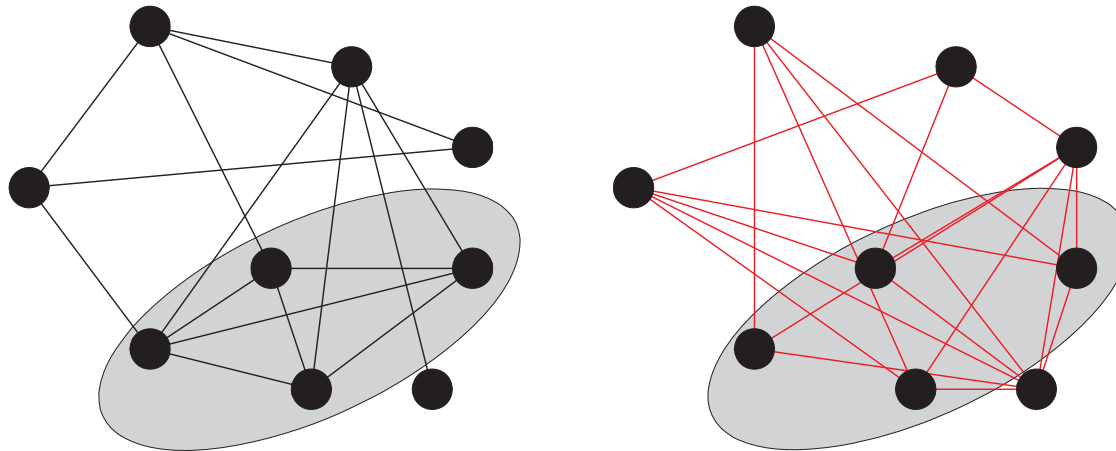
CLIQUE (concluded)



CLIQUE Is NP-Complete^a

Corollary 47 *CLIQUE is NP-complete.*

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



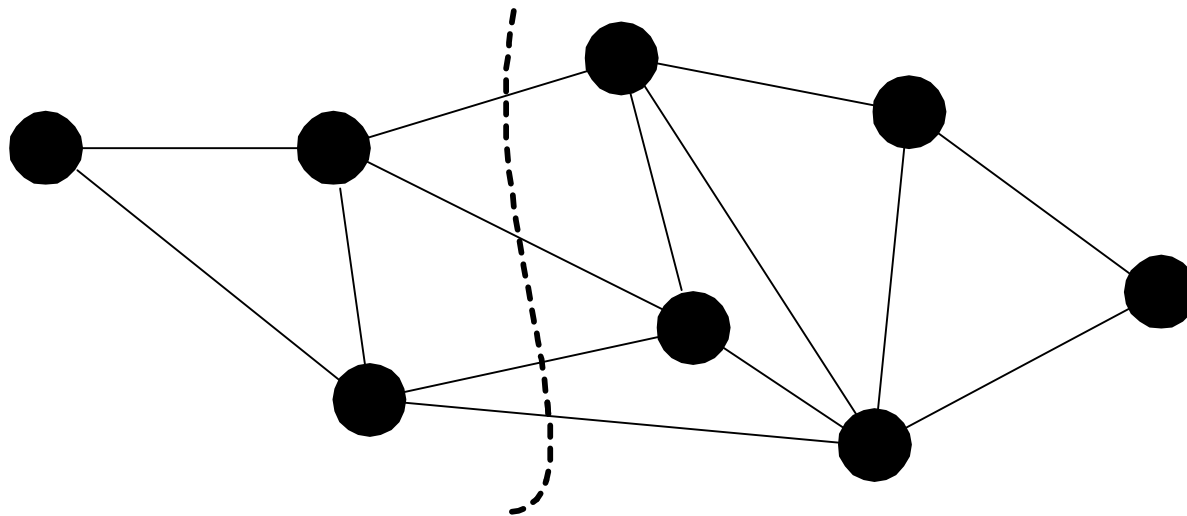
^aKarp (1972).

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 asks for the minimum cut size.
- MIN CUT $\in \text{P}$ by the maxflow algorithm.^a
- MAX CUT asks if there is a cut of size at least K .
 - K is part of the input.

^aFord & Fulkerson (1962); Orlin (2012) improves the running time to $O(|V| \cdot |E|)$.

A Cut of Size 4



MIN CUT and MAX CUT (concluded)

- MAX CUT has applications in circuit layout.
 - The minimum area of a VLSI layout of a graph is not less than the square of its maximum cut size.^a

^aRaspaud, Sýkora, & Vrřto (1995); Mak & Wong (2000).

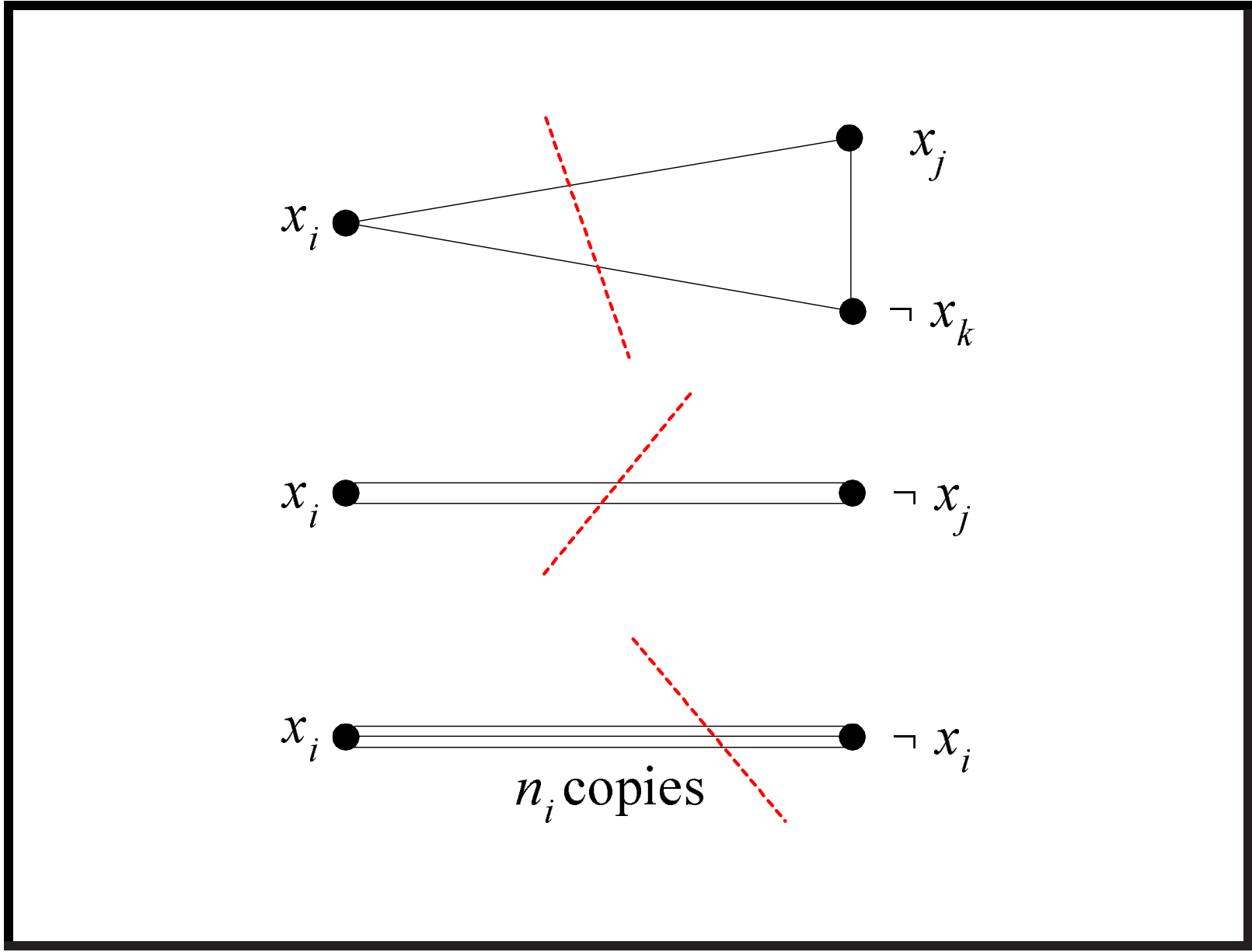
MAX CUT Is NP-Complete^a

- We will reduce NAESAT to MAX CUT.
- Given a 3SAT formula ϕ with m clauses, we shall construct a graph $G = (V, E)$ and a goal K .
- Furthermore, there is a cut of size at least K if and only if ϕ is NAE-satisfiable.
- Our graph will have *multiple* edges between two nodes.
 - Each such edge contributes one to the cut if its nodes are separated.

^aKarp (1972); Garey, Johnson, & Stockmeyer (1976). MAX CUT remains NP-complete even for graphs with maximum degree 3 (Makedon, Papadimitriou, & Sudborough, 1985).

The Proof

- Suppose ϕ 's m clauses are C_1, C_2, \dots, C_m .
- The boolean variables are x_1, x_2, \dots, x_n .
- G has $2n$ nodes: $x_1, x_2, \dots, x_n, \neg x_1, \neg x_2, \dots, \neg x_n$.
- Each clause with 3 distinct literals makes a triangle in G .
- For each clause with two identical literals, there are two parallel edges between the two distinct literals.
 - Call it a degenerate triangle.



The Proof (continued)

- Assume ϕ has no clauses with only one distinct literal (why?).
- Ignore clauses containing two opposite literals x_i and $\neg x_i$ (why?).
- For each variable x_i , add n_i copies of edge $[x_i, \neg x_i]$, where n_i is the number of occurrences of x_i and $\neg x_i$ in ϕ .
- Note that

$$\sum_{i=1}^n n_i = 3m.$$

- The summation counts the number of literals in ϕ .

Take

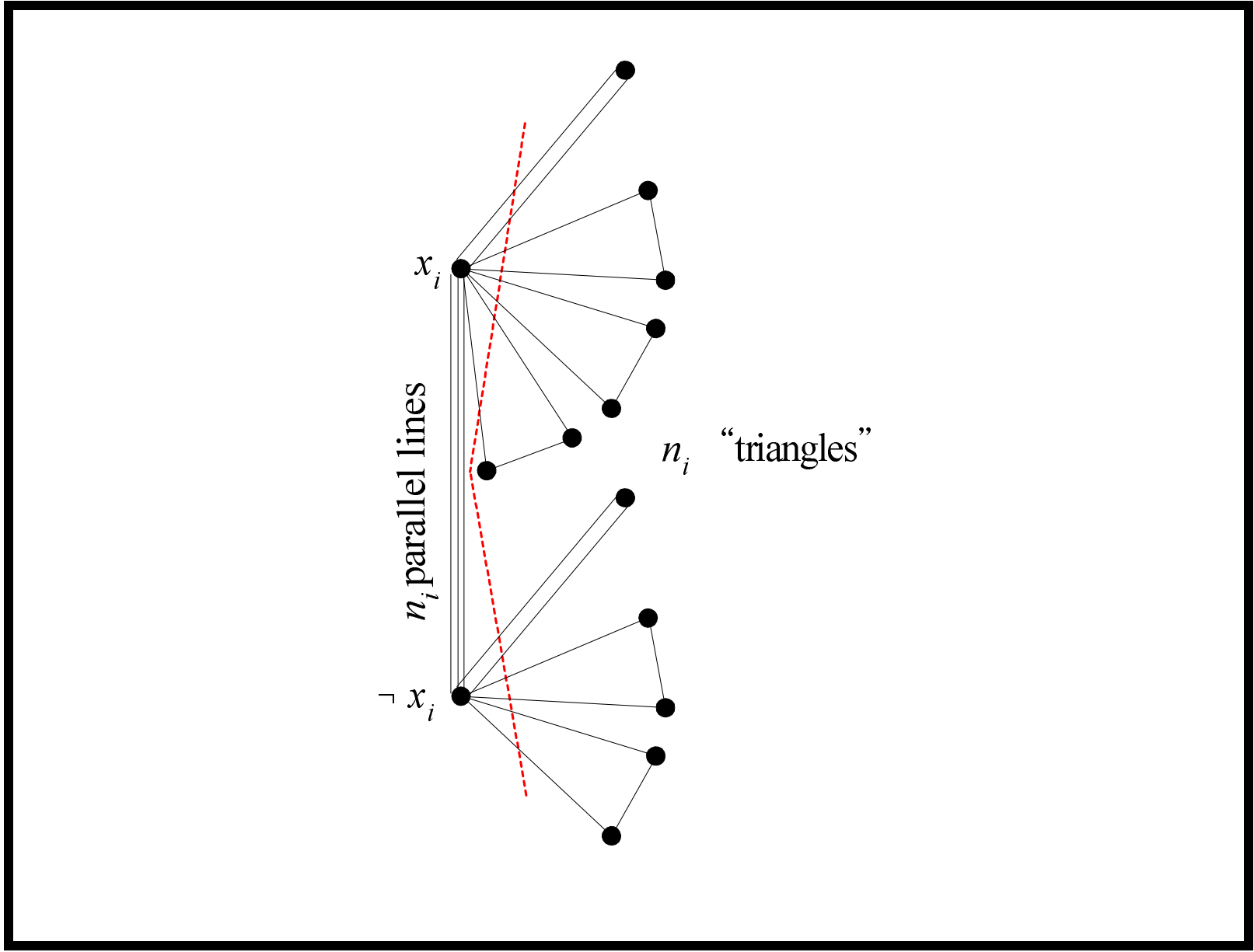
$$(x_1 \vee x_2 \vee x_2) \wedge (x_1 \vee \neg x_3 \vee \neg x_3) \wedge (\neg x_1 \vee \neg x_2 \vee x_3).$$

- Then $n_1 = n_2 = n_3 = 3$

The Proof (continued)

- Set $K = 5m$.
- Suppose there is a cut $(S, V - S)$ of size $5m$ or more.
- A clause (a triangle, i.e.) contributes at most 2 to a cut however you split it.^a
- Suppose some x_i and $\neg x_i$ are on the *same* side of the cut.
- They *together* contribute at most $2n_i$ edges to the cut.
 - They appear in at most n_i different *clauses*.
 - A clause contributes at most 2 to a cut.

^aSee p. 402.



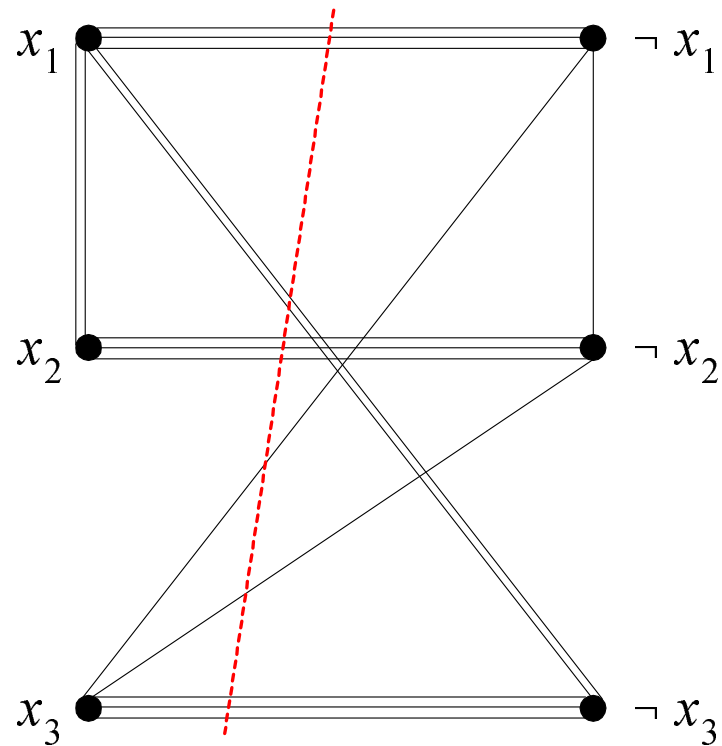
The Proof (continued)

- Either x_i or $\neg x_i$ contributes at most n_i to the cut by the pigeonhole principle.
- Changing the side of that literal does *not decrease* the size of the cut.
- Hence we assume variables are separated from their negations.
- The total number of edges in the cut that join opposite literals x_i and $\neg x_i$ is $\sum_{i=1}^n n_i$.
- But we knew $\sum_{i=1}^n n_i = 3m$.

The Proof (concluded)

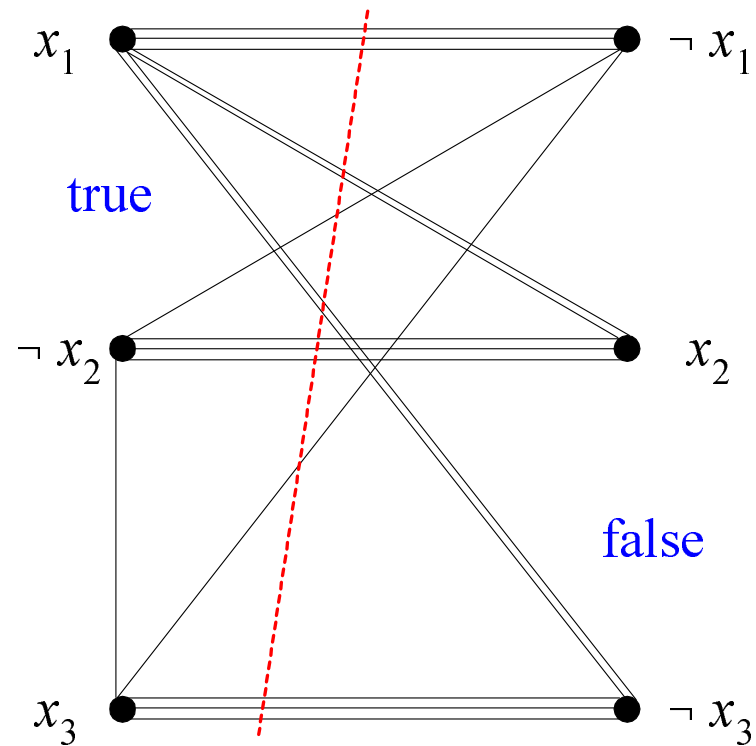
- The *remaining* $K - 3m \geq 2m$ edges in the cut must come from the m triangles that correspond to clauses.
- Each can contribute at most 2 to the cut.
- So all are split.
- A split clause means at least one of its literals is true and at least one false.
- The other direction is left as an exercise.

This Cut Does Not Meet the Goal $K = 5 \times 3 = 15$



- $(x_1 \vee x_2 \vee x_2) \wedge (x_1 \vee \neg x_3 \vee \neg x_3) \wedge (\neg x_1 \vee \neg x_2 \vee x_3)$.
- The cut size is $13 < 15$.

This Cut Meets the Goal $K = 5 \times 3 = 15$



- $(x_1 \vee x_2 \vee x_2) \wedge (x_1 \vee \neg x_3 \vee \neg x_3) \wedge (\neg x_1 \vee \neg x_2 \vee x_3)$.
- The cut size is now 15.

Remarks

- We had proved that MAX CUT is NP-complete for multigraphs.
- How about proving the same thing for simple graphs?^a
- How to modify the proof to reduce 4SAT to MAX CUT?^b
- All NP-complete problems are mutually reducible by definition.^c
 - So they are equally hard in this sense.^d

^aContributed by Mr. Tai-Dai Chou (J93922005) on June 2, 2005.

^bContributed by Mr. Chien-Lin Chen (J94922015) on June 8, 2006.

^cContributed by Mr. Ren-Shuo Liu (D98922016) on October 27, 2009.

^dContributed by Mr. Ren-Shuo Liu (D98922016) on October 27, 2009.

MAX BISECTION

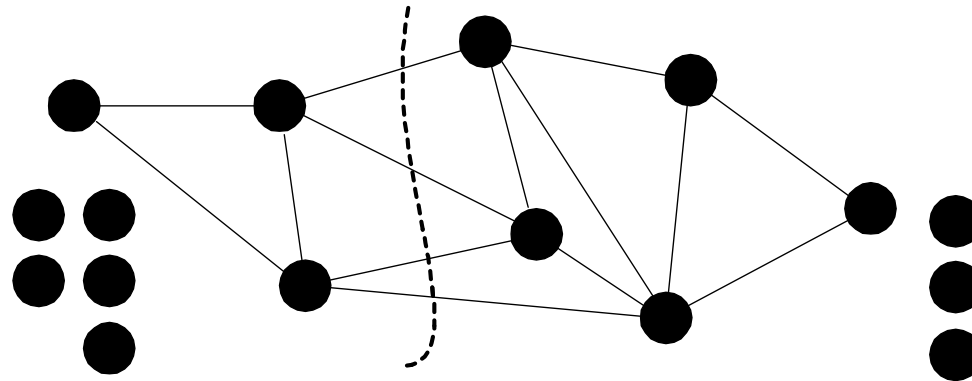
- MAX CUT becomes MAX BISECTION if we require that $|S| = |V - S|$.
- It has many applications, especially in VLSI layout.

MAX BISECTION Is NP-Complete

- We shall reduce the more general MAX CUT to MAX BISECTION.
- Add $|V| = n$ **isolated nodes** to G to yield G' .
- G' has $2n$ nodes.
- G' 's goal K is identical to G 's
 - As the new nodes have no edges, they contribute 0 to the cut.
- This completes the reduction.

The Proof (concluded)

- A cut $(S, V - S)$ can be made into a bisection by allocating the new nodes between S and $V - S$.
- Hence each cut of G can be made a cut of G' of the same size, and vice versa.



BISECTION WIDTH

- BISECTION WIDTH is like MAX BISECTION except that it asks if there is a bisection of size *at most* K (sort of MIN BISECTION).
- Unlike MIN CUT, BISECTION WIDTH is NP-complete.
- We reduce MAX BISECTION to BISECTION WIDTH.
- Given a graph $G = (V, E)$, where $|V|$ is even, we generate the complement^a of G .
- Given a goal of K , we generate a goal of $n^2 - K$.^b

^aRecall p. 396.

^b $|V| = 2n$.

The Proof (concluded)

- To show the reduction works, simply notice the following easily verifiable claims.
 - A graph $G = (V, E)$, where $|V| = 2n$, has a bisection of size K if and only if the complement of G has a bisection of size $n^2 - K$.
 - So G has a bisection of size $\geq K$ if and only if its complement has a bisection of size $\leq n^2 - K$.

HAMILTONIAN PATH Is NP-Complete^a

Theorem 48 *Given an undirected graph, the question whether it has a Hamiltonian path is NP-complete.*

^aKarp (1972).

A Hamiltonian Path at IKEA, Covina, California?



Random HAMILTONIAN CYCLE

- Consider a random graph where each pair of nodes are connected by an edge independently with probability $1/2$.
- Then it contains a Hamiltonian cycle with probability $1 - o(1)$.^a

^aFrieze & Reed (1998).

TSP (D) Is NP-Complete

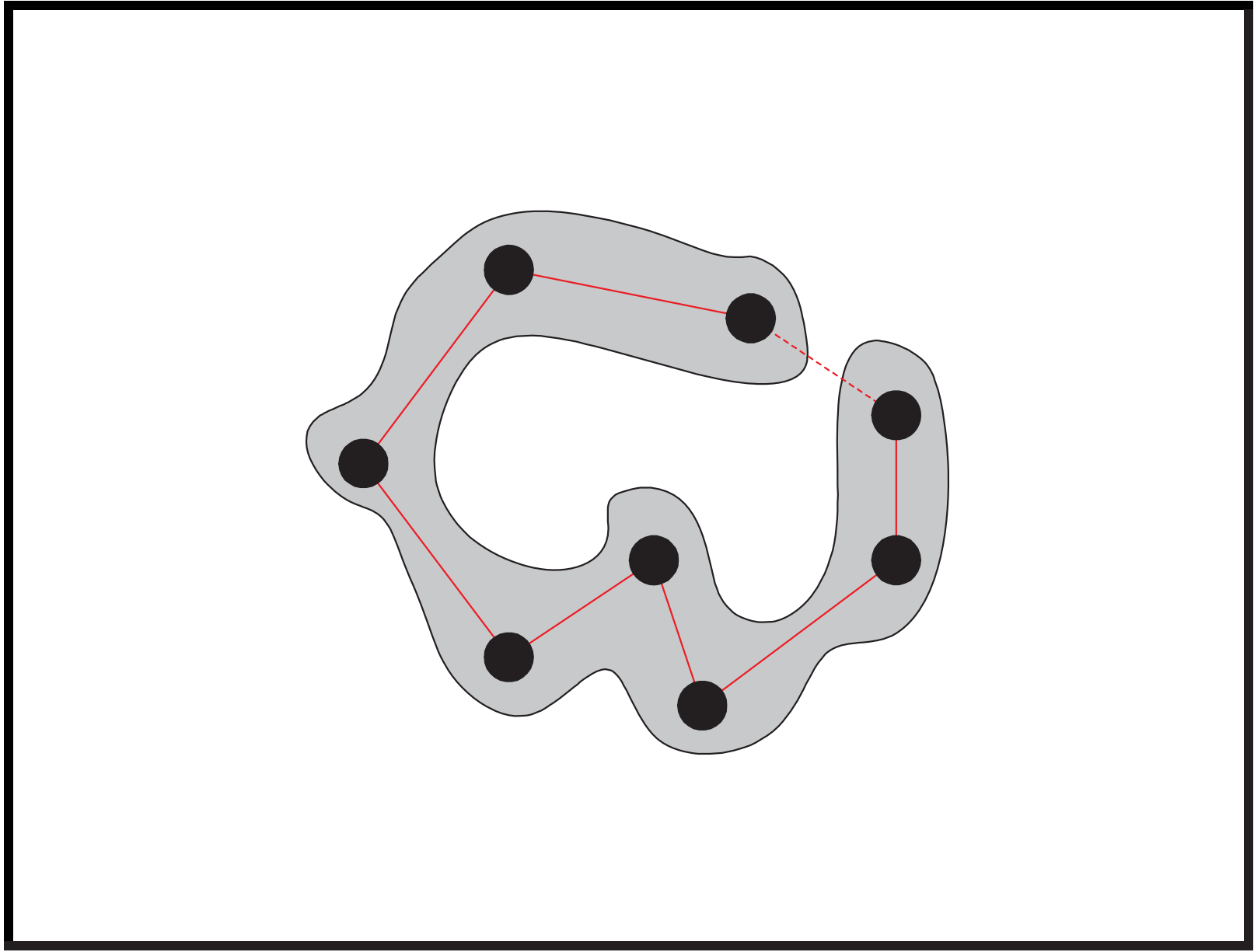
Corollary 49 TSP (D) *is NP-complete.*

- We will reduce HAMILTONIAN PATH to TSP (D).
- Consider a graph G with n nodes.
- Create a weighted complete graph G' with the same nodes as G .
- Set $d_{ij} = 1$ on G' if $[i, j] \in G$ and $d_{ij} = 2$ on G' if $[i, j] \notin G$.
 - Note that G' is a complete graph.
- Set the budget $B = n + 1$.
- This completes the reduction.

TSP (D) Is NP-Complete (continued)

- Suppose G' has a tour^a of distance at most $n + 1$.
- Then that tour on G' must contain at most one edge with weight 2.
- If a tour on G' contains one edge with weight 2, remove that edge to arrive at a Hamiltonian path for G .
- Suppose a tour on G' contains no edge with weight 2.
- Remove any edge to arrive at a Hamiltonian path for G .

^aA tour is a cycle, not a path.



TSP (D) Is NP-Complete (concluded)

- On the other hand, suppose G has a Hamiltonian path.
- There is a tour on G' containing at most one edge with weight 2.
 - Start with a Hamiltonian path.
 - Insert the edge connecting the beginning and ending nodes to yield a tour.
- The total cost is then at most $(n - 1) + 2 = n + 1 = B$.
- We conclude that there is a tour of length B or less on G' if and only if G has a Hamiltonian path.

Random TSP

- Suppose each distance d_{ij} is picked uniformly and independently from the interval $[0, 1]$.
- Then the total distance of the shortest tour has a mean value of $\beta\sqrt{n}$ for some positive β .^a
- In fact, the total distance of the shortest tour deviates from the mean by more than t with probability at most $e^{-t^2/(4n)}$.^b

^aBeardwood, Halton, & Hammersley (1959).

^bRhee & Talagrand (1987); Dubhashi & Panconesi (2012).

Graph Coloring

- k -COLORING: Can the nodes of a graph be colored with $\leq k$ colors such that no two adjacent nodes have the same color?^a
- 2-COLORING is in P (why?).
- But 3-COLORING is NP-complete (see next page).

^a k is *not* part of the input; k is part of the problem statement.

3-COLORING Is NP-Complete^a

- We will reduce NAESAT to 3-COLORING.
- We are given a set of clauses C_1, C_2, \dots, C_m each with 3 literals.
- The boolean variables are x_1, x_2, \dots, x_n .
- We now construct a graph that can be colored with colors $\{0, 1, 2\}$ if and only if all the clauses can be NAE-satisfied.

^aKarp (1972).

The Proof (continued)

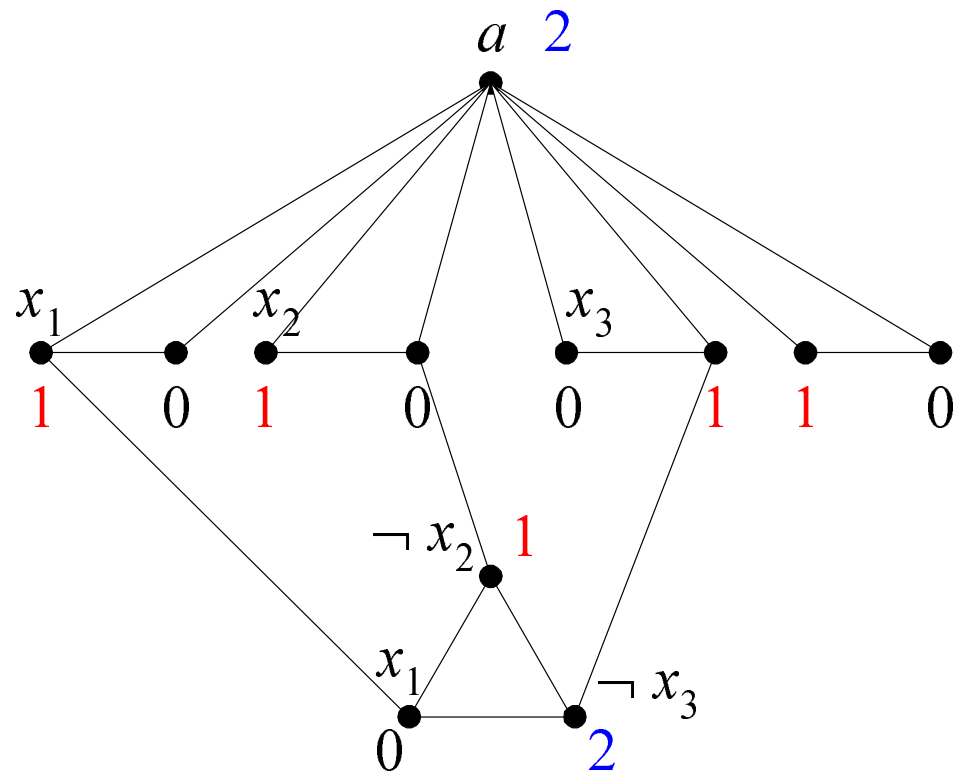
- Every variable x_i is involved in a triangle $[a, x_i, \neg x_i]$ with a common node a .
- Each clause $C_i = (c_{i1} \vee c_{i2} \vee c_{i3})$ is also represented by a triangle

$$[c_{i1}, c_{i2}, c_{i3}].$$

- Node c_{ij} and a node in an a -triangle $[a, x_k, \neg x_k]$ with the same label represent *distinct* nodes.
- There is an edge between literal c_{ij} in the a -triangle and the node representing the j th literal of C_i .^a

^aAlternative proof: There is an edge between $\neg c_{ij}$ and the node that represents the j th literal of C_i . Contributed by Mr. Ren-Shuo Liu (D98922016) on October 27, 2009.

Construction for $\dots \wedge (x_1 \vee \neg x_2 \vee \neg x_3) \wedge \dots$



The Proof (continued)

Suppose the graph is 3-colorable.

- Assume without loss of generality that node a takes the color 2.
- A triangle must use up all 3 colors.
- As a result, one of x_i and $\neg x_i$ must take the color 0 and the other 1.

The Proof (continued)

- Treat 1 as **true** and 0 as **false**.^a
 - We are dealing with the a -triangles here, not the clause triangles yet.
- The resulting truth assignment is clearly contradiction free.
- As each clause triangle contains one color 1 and one color 0, the clauses are NAE-satisfied.
 - Here, treat 0 as **true** and 1 as **false**.
 - Ignore 2's truth value as it is irrelevant now.

^aThe opposite also works.

The Proof (continued)

Suppose the clauses are NAE-satisfiable.

- For each a -triangle:
 - Color node a with color 2.
 - Color the nodes representing literals by their truth values (color 0 for **false** and color 1 for **true**).

The Proof (continued)

- For each clause triangle:
 - Pick any two literals with opposite truth values.^a
 - Color the corresponding nodes with 0 if the literal is true and 1 if it is false.
 - Color the remaining node with color 2 regardless of its truth value.

^aBreak ties arbitrarily.

The Proof (concluded)

- The coloring is legitimate.
 - If literal w of a clause triangle has color 2, then its color will never be an issue.
 - If literal w of a clause triangle has color 1, then it must be connected up to literal w with color 0.
 - If literal w of a clause triangle has color 0, then it must be connected up to literal w with color 1.

More on 3-COLORING and the Chromatic Number

- 3-COLORING remains NP-complete for planar graphs.^a
- Assume G is 3-colorable.
- There is a classic algorithm that finds a 3-coloring in time $O(3^{n/3}) = 1.4422^n$.^b
- It can be improved to $O(1.3289^n)$.^c

^aGarey, Johnson, & Stockmeyer (1976); Dailey (1980).

^bLawler (1976).

^cBeigel & Eppstein (2000).

More on 3-COLORING and the Chromatic Number (concluded)

- The **chromatic number** $\chi(G)$ is the smallest number of colors needed to color a graph G .
- There is an algorithm to find $\chi(G)$ in time $O((4/3)^{n/3}) = 2.4422^n$.^a
- It can be improved to $O((4/3 + 3^{4/3}/4)^n) = O(2.4150^n)$ ^b and $2^n n^{O(1)}$.^c
- Computing $\chi(G)$ cannot be easier than 3-COLORING.^d

^aLawler (1976).

^bEppstein (2003).

^cKoivisto (2006).

^dContributed by Mr. Ching-Hua Yu (D00921025) on October 30, 2012.