# Analysis of Algorithms CS 477/677

Instructor: Monica Nicolescu Lecture 26

### NP-Completeness

- Polynomial-time algorithms
  - on inputs of size n, worst-case running time is  $O(n^k)$ , for a constant k
- Not all problems can be solved in polynomial time
  - Some problems cannot be solved by any computer no matter how much time is provided (Turing's Halting problem) – such problems are called **undecidable**
  - Some problems can be solved but not in O(n<sup>k</sup>)

#### Class of "P" Problems

 Class P consists of (decision) problems that are solvable in polynomial time:

there exists an algorithm that can solve the problem in  $O(n^k)$ , k constant

- Problems in P are also called tractable
- Problems not in P are also called intractable
  - Can be solved in reasonable time only for small inputs

# Optimization & Decision Problems

#### Decision problems

Given an input and a question regarding a problem,
 determine if the answer is yes or no

#### Optimization problems

- Find a solution with the "best" value
- Optimization problems can be cast as decision problems that are easier to study
  - E.g.: Shortest path: G = unweighted directed graph
    - Find a path between u and v that uses the fewest edges
    - Does a path exist from u to v consisting of at most k edges?

### Nondeterministic Algorithms

# **Nondeterministic algorithm** = two stage procedure:

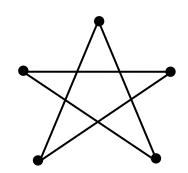
- 1) Nondeterministic ("guessing") stage:
  - generate an arbitrary string that can be thought of as a candidate solution ("certificate")
- 2) Deterministic ("verification") stage:
  - take the certificate and the instance to the problem and return YES if the certificate represents a solution
- Nondeterministic polynomial (NP) = verification stage is polynomial

#### Class of "NP" Problems

- Class NP consists of problems that are verifiable in polynomial time (i.e., could be solved by nondeterministic polynomial algorithms)
  - If we were given a "certificate" of a solution, we could verify that the certificate is correct in time polynomial to the size of the input

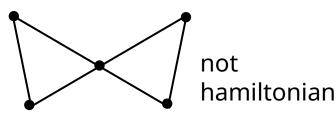
### E.g.: Hamiltonian Cycle

- Given: a directed graph G = (V, E), determine a simple cycle that contains each vertex in V
  - Each vertex can only be visited once
- Certificate:
  - Sequence:  $\langle v_1, v_2, v_3, ..., v_{|V|} \rangle$

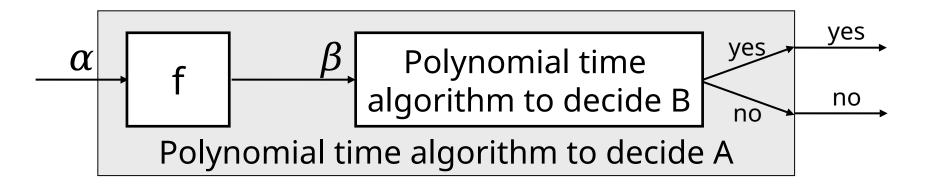


hamiltonian

- Verification:
  - $(v_i, v_{i+1}) \in E \text{ for } i = 1, ..., |V|$   $(v_{|V|}, v_1) \in E$
  - 2)



### Polynomial Reduction Algorithm



- To solve a decision problem A in polynomial time
  - 1. Use a polynomial time reduction algorithm to transform A into B
  - 2. Run a known polynomial time algorithm for B
  - 3. Use the answer for B as the answer for A CS 477/677 Lecture 26

#### Reductions

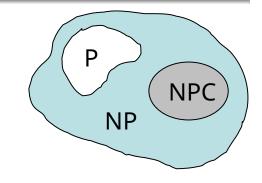
Given two problems A, B, we say that A is

**reducible** to B (A 
$$\leq_p$$
 B) if:

- 1. There exists a function f that converts the input of A to an input of B in polynomial time
- 2.  $A(i) = YES \iff B(f(i)) = YES$  (for every input i)

### NP-Completeness

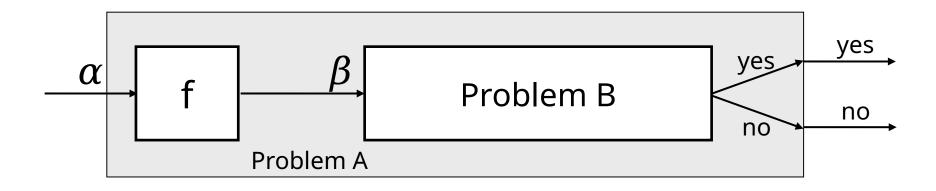
A problem B is NP-complete (NPC) if:



2) 
$$A \leq_p B$$
 for all  $A \in \mathbf{NP}$ 

- If B satisfies only property 2) we say that B is NP-hard
- No polynomial time algorithm has been discovered for an NP-Complete problem
- No one has ever proven that no polynomial time algorithm can exist for any NP-Complete problem

### Reduction and NP-Completeness



- Suppose we know:
  - No polynomial time algorithm exists for problem A
  - We have a polynomial reduction f from A to B
- ⇒ No polynomial time algorithm exists for B

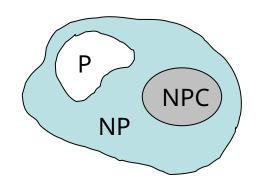
### Proving NP-Completeness

Theorem: If A is NP-Complete and A  $\leq_{D}$  B

⇒ B is NP-Hard

In addition, if  $B \in NP$ 

⇒ B is NP-Complete



**Proof**: Assume that  $B \in P$ 

Since  $A \leq_p B \Rightarrow A \in P$  contradiction, so  $B \notin P$ 

If B  $\in$  NP  $\Rightarrow$  B  $\in$  NP-Complete (by definition of NP-C)

If B  $\notin$  NP  $\Rightarrow$  B  $\in$  NP-Hard (by definition of NP-H)

### Proving NP-Completeness

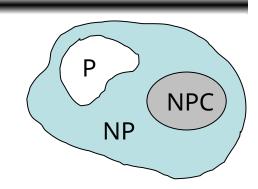
#### 1. Prove that the problem B is in NP

- A randomly generated string can be checked in polynomial time to determine if it represents a solution
- 2. Show that **one known** NP-Complete problem can be transformed to B in polynomial time
  - No need to check that all NP-Complete problems are reducible to B

#### Is P = NP?

Any problem in P is also in NP:

$$P \subseteq NP$$



- We can solve problems in P, even without having a certificate
- The big (and open question) is whether P = NP

Theorem: If any NP-Complete problem can be solved in polynomial time  $\Rightarrow$  then P = NP.

### P & NP-Complete Problems

#### Shortest simple path

- Given a graph G = (V, E) find a **shortest** path from a source to all other vertices
- Polynomial solution: O(VE)

#### Longest simple path

- Given a graph G = (V, E) find a **longest** path from a
   source to all other vertices
- NP-complete

### P & NP-Complete Problems

#### Euler tour

- Given G = (V, E) a connected, directed graph, find a cycle that traverses each edge of G exactly once (may visit a vertex multiple times)
- Polynomial solution O(E)

#### Hamiltonian cycle

- G = (V, E) a connected, directed graph find a cycle
   that visits each vertex of G exactly once
- NP-complete

### Boolean Formula Satisfiability

# Formula Satisfiability Problem: a boolean formula **@** composed of

- 1. n boolean variables: x<sub>1</sub>, x<sub>2</sub>, ..., x<sub>n</sub>
- 2. m boolean connectives: Λ (AND), V (OR), ¬ (NOT), → (implication), ←→ (equivalence, "if and only if")
- 3. Parentheses

**Satisfying assignment:** an assignment of values (0, 1) to variables  $x_i$  that causes  $\Phi$  to evaluate to 1

E.g.: 
$$\Phi = (x_1 \ V \ x_2) \ \Lambda (x_1 \ V \ \neg x_2) \ \Lambda (\neg x_1 \ V \ \neg x_2)$$
  
Certificate:  $x_1 = 1, x_2 = 0 \Rightarrow \Phi = 1 \ \Lambda \ 1 \ \Lambda \ 1 = 1$ 

Formula Satisfiability is first to be proven NP-Complete
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### 3-CNF Satisfiability

## 3-CNF (clause normal form) Satisfiability Problem:

- n boolean variables: x<sub>1</sub>, x<sub>2</sub>, ..., x<sub>n</sub>
- **Literal**:  $x_i$  or  $\neg x_i$  (a variable or its negation)
- Clause: c<sub>i</sub> = an OR of three literals
- Formula:  $\Phi = c_1 \wedge c_2 \wedge ... \wedge c_m$  (m clauses)
- E.g.:

$$\mathbf{\Phi} = (\mathbf{X}_1 \ \mathbf{V} \ \neg \mathbf{X}_1 \ \mathbf{V} \ \neg \mathbf{X}_2) \ \Lambda \ (\mathbf{X}_3 \ \mathbf{V} \ \mathbf{X}_2 \ \mathbf{V} \ \mathbf{X}_4) \ \Lambda$$
$$(\neg \mathbf{X}_1 \ \mathbf{V} \ \neg \mathbf{X}_3 \ \mathbf{V} \ \neg \mathbf{X}_4)$$

3-CNF is NP-Complete

### Clique

#### **Clique Problem:**

- Undirected graph G = (V, E)
- Clique: a subset of vertices in V all connected to each other by edges in E (i.e., forming a complete graph)

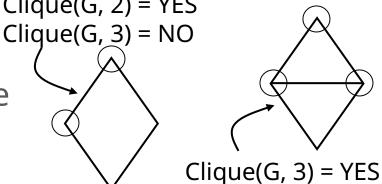
 Size of a clique: number of vertices it contains Clique(G, 2) = YES

#### **Optimization problem:**

Find a clique of maximum size

#### **Decision problem:**

– Does G have a clique of size k?



Clique(G, 4) = NO

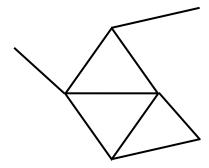
### Clique Verifier

- **Given**: an undirected graph G = (V, E)
- Problem: Does G have a clique of size k?
- Certificate:
  - A set of k nodes





 Let's prove that the clique problem is NP-Complete



### 3-CNF $\leq_p$ Clique

Start with an instance of 3-CNF:

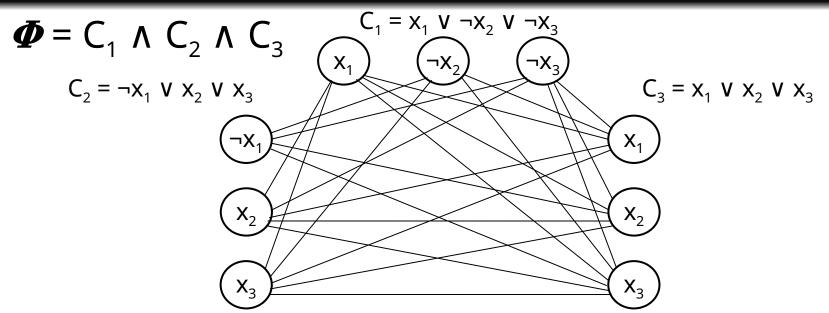
$$-\Phi = C_1 \wedge C_2 \wedge ... \wedge C_k$$
(k clauses)

- Each clause  $C_r$  has three literals:  $C_r = I_1^r v I_2^r v I_3^r$ 

#### • Idea:

– Construct a graph G such that  $\Phi$  is satisfiable if and only if G has a clique of size k

### 3-CNF $\leq_p$ Clique

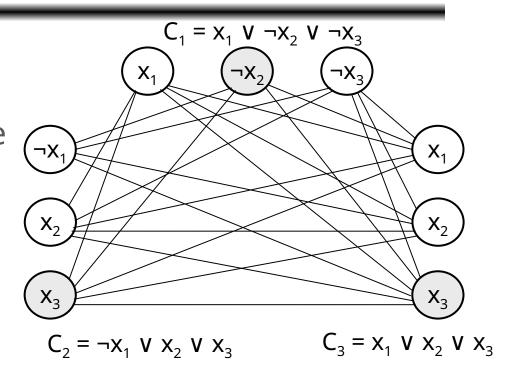


- For each clause  $C_r = I_1^r v I_2^r v I_3^r$  place a triple of vertices  $v_1^r$ ,  $v_2^r$ ,  $v_3^r$  in V
- Put an edge between two vertices v<sub>i</sub><sup>r</sup> and v<sub>i</sub><sup>s</sup> if:
  - v<sub>i</sub><sup>r</sup> and v<sub>i</sub><sup>s</sup> are in different triples
  - $I_i^r$  is not the negation of  $I_j^s$

### 3-CNF $\leq_p$ Clique

$$\Phi = C_1 \wedge C_2 \wedge C_3$$

- Suppose **Φ** has a satisfying assignment
  - Each clause C<sub>r</sub> has some
     literal assigned to 1 –
     this corresponds to a
     vertex v<sub>i</sub><sup>r</sup>
  - Picking one such literal from each C<sub>r</sub> ⇒ a set V'

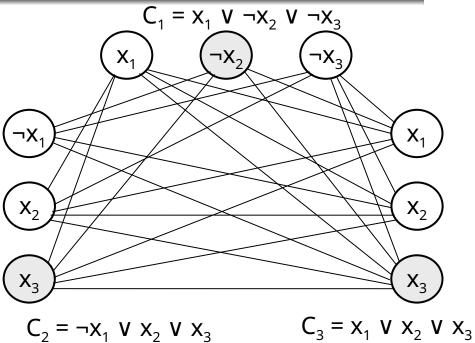


- Claim: Vitices clique
  - $-\forall v_i^r, v_j^s \in V'$  the corresponding literals are 1 ⇒ cannot be complements
  - by the design of G the edge  $(v_i^r, v_j^s)$  ∈ E

### 3-CNF ≤<sub>p</sub> Clique

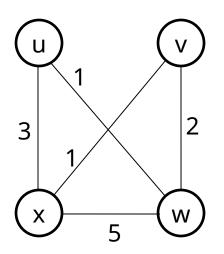
$$\Phi = C_1 \wedge C_2 \wedge C_3$$

- Suppose G has a clique of size k
  - No edges between nodes in the same clause
  - Clique contains only one vertex from each clause
  - Assign 1 to vertices in C<sub>2</sub> ¬x<sub>1</sub> v x<sub>2</sub> v x<sub>3</sub> the clique (we can do it because the literals of these vertices cannot belong to complementary literals)
  - Each clause is satisfied  $\Rightarrow \Phi$  is satisfied



### The Traveling Salesman Problem

- G = (V, E), |V| = n, vertices
   represent cities
- Cost: c(i, j) = cost of travel from city i to city j
- Problem: salesman should make a tour (hamiltonian cycle):
  - Visit each city only once
  - Finish at the city he started from
  - Total cost is minimum
- TSP = tour with cost at most k



 $\langle u, w, v, x \rangle$ 

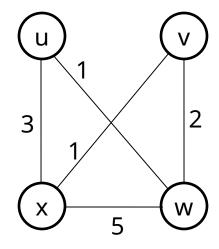
#### $TSP \in NP$

#### Certificate:

- Sequence of n vertices, cost
- E.g.:  $\langle u, w, v, x \rangle$ , 7

#### Verification:

- Each vertex occurs only once
- Sum of costs is at most k



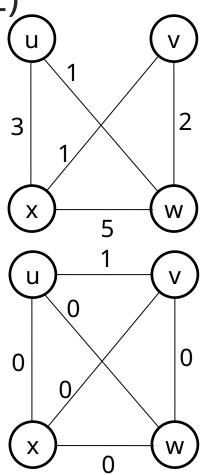
### $\mathsf{HAM}\text{-}\mathsf{CYCLE} \leq_{\mathsf{p}} \mathsf{TSP}$

- Start with a Hamiltonian cycle G = (V, E)
- Form the complete graph G' = (V, E')

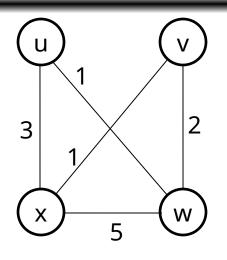
$$E' = \{(i, j): i, j \in V \text{ and } i \neq j\}$$

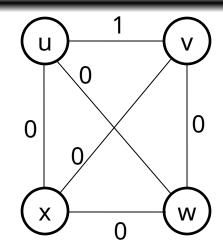
$$c(i, j) = \begin{cases} 0 & \text{if } (i, j) \in E \\ 1 & \text{if } (i, j) \notin E \end{cases}$$

- Let's prove that:
- G has a hamiltonian cycle ⇔
   G' has a tour of cost at most 0



### $\mathsf{HAM}\text{-}\mathsf{CYCLE} \leq_{\mathsf{p}} \mathsf{TSP}$





- G has a hamiltonian cycle h
  - $\Rightarrow$  Each edge in h  $\in$  E  $\Rightarrow$  has cost 0 in G'
  - ⇒ h is a tour in G' with cost 0
- G' has a tour h' of cost at most 0
  - ⇒ Each edge on tour must have cost 0
  - ⇒ h' contains only edges in E

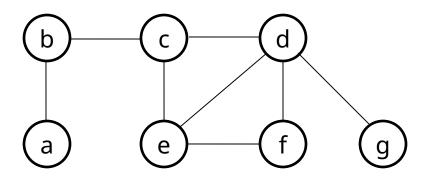
### **Approximation Algorithms**

Various ways to get around NP-completeness:

- 1. If inputs are small, an algorithm with exponential time may be satisfactory
- 2. Isolate special cases, solvable in polynomial time
- 3. Find near-optimal solutions in polynomial time
  - Approximation algorithms
  - Local search (hill climbing)

#### The Vertex-Cover Problem

- Vertex cover of G = (V, E), undirected graph
  - A subset  $V' \subseteq V$  that covers all the edges in G

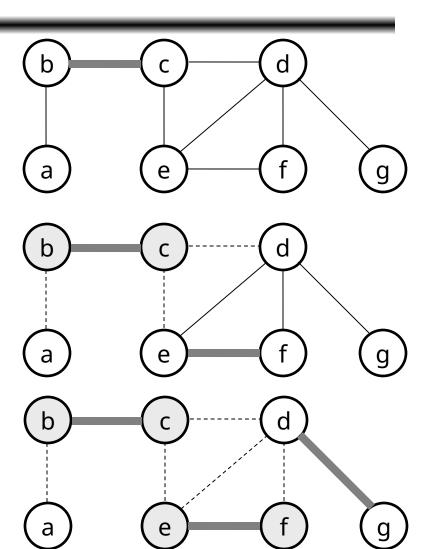


#### Approximate solution (greedy):

- Start with a list of all edges
- Repeatedly pick an arbitrary edge (u, v)
- Add its endpoints u and v to the vertex-cover set
- Remove from the list all edges incident on u or v

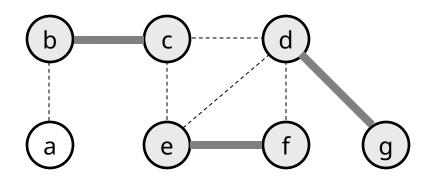
### APPROX-VERTEX-COVER(G)

- 1. C ← Ø
- 2. E' ← E[G]
- 3. while  $E' \neq \emptyset$
- 4. **do** choose (u, v) arbitrary from E'
- 5.  $C \leftarrow C \bigcup \{u, v\}$
- 6. remove from E' all edges incident on u, v
- **7.** return C

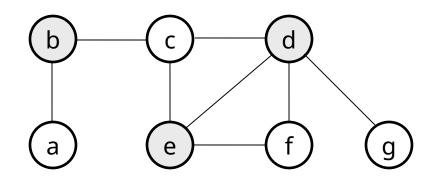


### APPROX-VERTEX-COVER(G)

APPROX-VERTEX-COVER:



Optimal VERTEX-COVER:



It can be proven that the approximation algorithm returns a solution that is no more than twice the optimal vertex cover.

### The Set Covering Problem

- Finite set X
- Family F of subsets of X:  $F = \{S_1, S_2, ..., S_n\}$

$$X = \bigcup_{S \in F} S$$

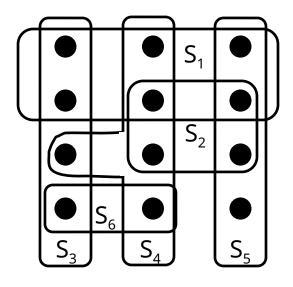
- Find a minimum-size subset C ⊆ F that covers all the elements in X
- Decision: given a number k find if there exist k sets  $S_{i1}$ ,  $S_{i2}$ , ...,  $S_{ik}$  such that:

$$S_{i1} \bigcup S_{i2} \bigcup ... \bigcup S_{ik} = X$$

### **Greedy Set Covering**

#### Idea:

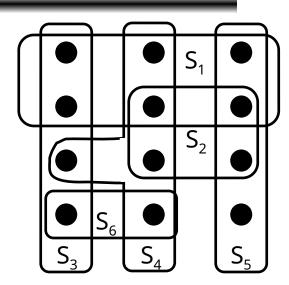
At each step pick a set S
 that covers the greatest
 number of remaining
 elements

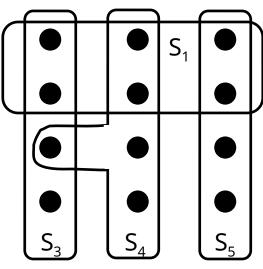


Optimal:  $C = \{S_3, S_4, S_5\}$ 

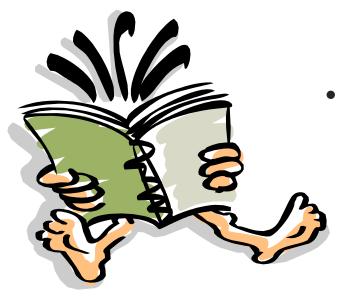
### GREEDY-SET-COVER(X, F)

- 1. U ← X
- 2. C ← Ø
- **3.** while ∪ **=**/∅
- 4. **do** select an  $S \in F$  that maximizes  $|S \cap U|$
- 5.  $U \leftarrow U S$
- 6.  $C \leftarrow C \cup \{S\}$
- 7. return C





### Readings



Chapters 25, 31

Optional, not required for final exam

### **ADDITIONAL PROOFS**

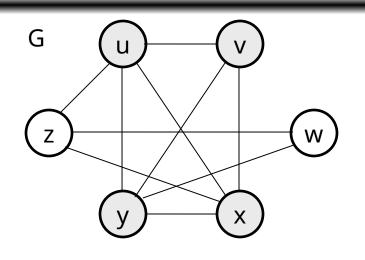
### Vertex Cover

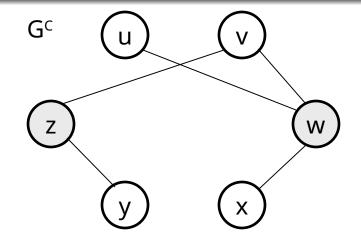
- G = (V, E), undirected graph
- Vertex cover = a subset V' ⊆ V
   which covers all the edges
  - if (u, v) ∈ E then u ∈ V' or v ∈ V' or both.
- **Size** of a vertex cover = number of vertices in it

#### **Problem:**

- Find a vertex cover of minimum size
- Does graph G have a vertex cover of size k?

# Clique ≤<sub>p</sub> Vertex Cover



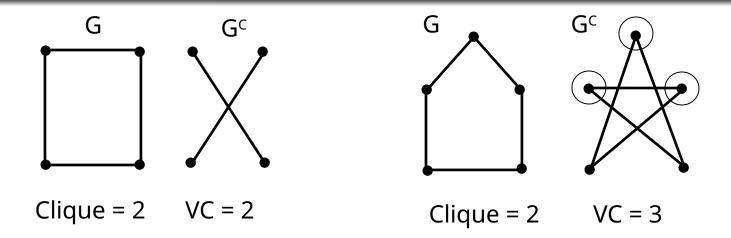


• G = (V, E)  $\Rightarrow$  complement graph G<sup>c</sup> = (V, E<sup>c</sup>) E<sup>c</sup> = {(u, v):, u, v  $\in$  V, and (u, v)  $\notin$  E}

#### Idea:

 $\langle G, k \rangle$  (clique)  $\rightarrow \langle G^c, |V|-k \rangle$  (vertex cover)

# Clique ≤<sub>p</sub> Vertex Cover (VC)

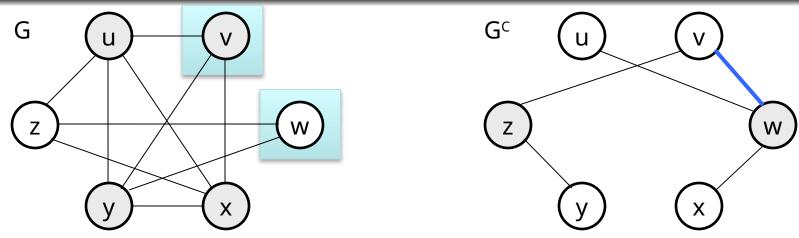


Size[Clique](G) + Size[Vertex Cover](G<sup>c</sup>) = n

- G has a clique of size k 

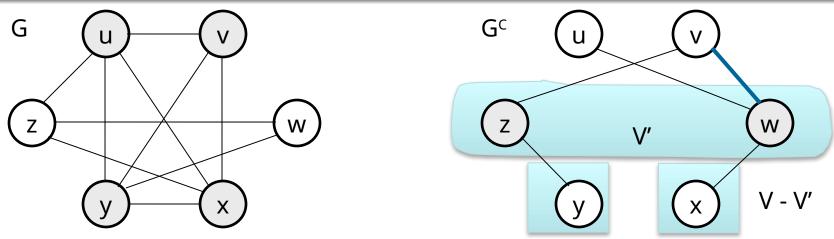
  G<sup>c</sup> has a vertex
  cover of size n − k
- S is a clique in  $G \iff V S$  is a vertex cover in  $G^c$

# Clique ≤<sub>p</sub> Vertex Cover



- Prove: G has a clique  $V' \subseteq V$ ,  $|V'| = k \Rightarrow V-V'$  is a VC in  $G^c$
- Let  $(v, w) \in E^c \Rightarrow (v, w) \notin E$ 
  - ⇒ v and w were not connected in E
  - ⇒ at least one of v or w does not belong in the clique V'
  - ⇒ at least one of v or w belongs in V V'
  - $\Rightarrow$  edge (v, w) is covered by V V'
  - $\Rightarrow$  edge (v, w) was arbitrary  $\Rightarrow$  every edge of E<sup>c</sup> is covered

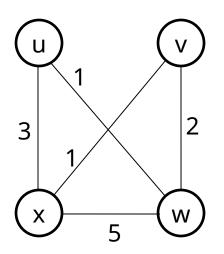
# Clique ≤<sub>p</sub> Vertex Cover



- Prove:  $G^c$  has a vertex cover  $V' \subseteq V$ ,  $|V'| = |V| k \Rightarrow V-V'$  is a clique in G
- For all  $v, w \in V$ , if  $(v, w) \in E^c$ 
  - $\Rightarrow$  v  $\in$  V' or w  $\in$  V' or both  $\in$  V'
    - $\Rightarrow$  For all x, y  $\in$  V, if x  $\notin$  V' and y  $\notin$  V':
    - $\Rightarrow$  no edge between x, y in E<sup>G</sup>  $\Rightarrow$  (x,y)  $\in$  E
      - $\Rightarrow$  V V' is a clique, of size |V| |V'| = k

### The Traveling Salesman Problem

- G = (V, E), |V| = n, vertices
   represent cities
- Cost: c(i, j) = cost of travel from city i to city j
- Problem: salesman should make a tour (hamiltonian cycle):
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 $\langle u, w, v, x \rangle$ 

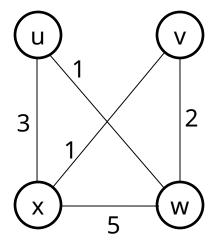
### $TSP \in NP$

#### Certificate:

- Sequence of n vertices, cost
- E.g.:  $\langle u, w, v, x \rangle$ , 7

#### Verification:

- Each vertex occurs only once
- Sum of costs is at most k



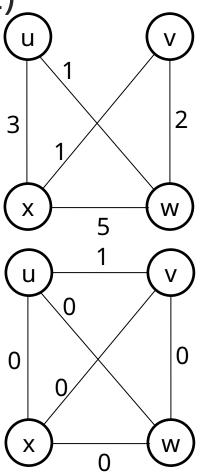
# $\mathsf{HAM}\text{-}\mathsf{CYCLE} \leq_{\mathsf{p}} \mathsf{TSP}$

- Start with a Hamiltonian cycle G = (V, E)
- Form the complete graph G' = (V, E')

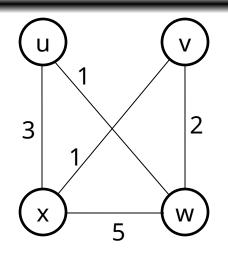
$$E' = \{(i, j): i, j \in V \text{ and } i \neq j\}$$

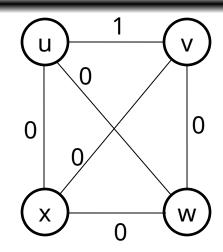
$$c(i, j) = \begin{cases} 0 & \text{if } (i, j) \in E \\ 1 & \text{if } (i, j) \notin E \end{cases}$$

- Let's prove that:



# $\mathsf{HAM}\text{-}\mathsf{CYCLE} \leq_{\mathsf{p}} \mathsf{TSP}$



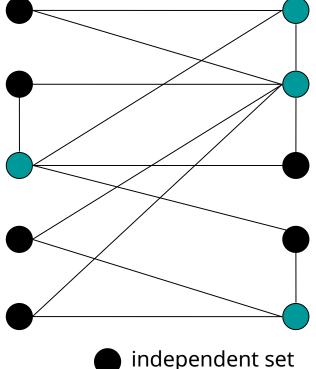


- G has a hamiltonian cycle h
  - $\Rightarrow$  Each edge in h  $\in$  E  $\Rightarrow$  has cost 0 in G'
  - ⇒ h is a tour in G' with cost 0
- G' has a tour h' of cost at most 0
  - ⇒ Each edge on tour must have cost 0
  - ⇒ h' contains only edges in E

### INDEPENDENT-SET

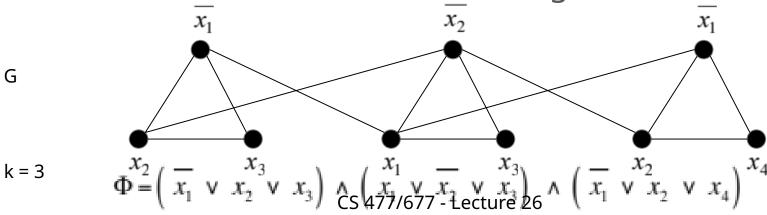
 Given a graph G = (V, E) and an integer k, is there a subset of vertices S ⊆ V such that |S| ≥ k, and for each edge at most one of its endpoints is in S?

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



## 3-CNF ≤<sub>p</sub> INDEPENDENT-SET

- Given an instance  $\Phi$  of 3-CNF, we construct an instance (G, k) of INDEPENDENT-SET that has an independent set of size k iff  $\Phi$  is satisfiable
- Construction
  - 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.



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# 3-CNF ≤<sub>p</sub> INDEPENDENT-SET

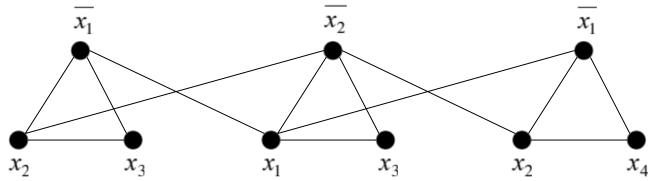
- Claim: G contains independent set of size  $k = |\Phi|$  iff  $\Phi$  is satisfiable
- Proof: "⇒" Let S be independent set of size k
  - S must contain exactly one vertex in each triangle
  - Set these literals to true
  - Truth assignment is consistent and all clauses are satisfied

 $x_1$   $x_2$   $x_1$   $x_2$   $x_3$   $x_4$   $x_4$   $x_5$   $x_4$ 

$$\Phi = \begin{pmatrix} \overline{x_1} & v & x_2 & v & x_3 \end{pmatrix} \wedge \begin{pmatrix} x_1 & v & \overline{x_2} & v & x_3 \end{pmatrix} \wedge \begin{pmatrix} \overline{x_1} & v & x_2 & v & x_4 \end{pmatrix}$$
CS 477/677 - Lecture 26

## 3-CNF ≤<sub>p</sub> INDEPENDENT-SET

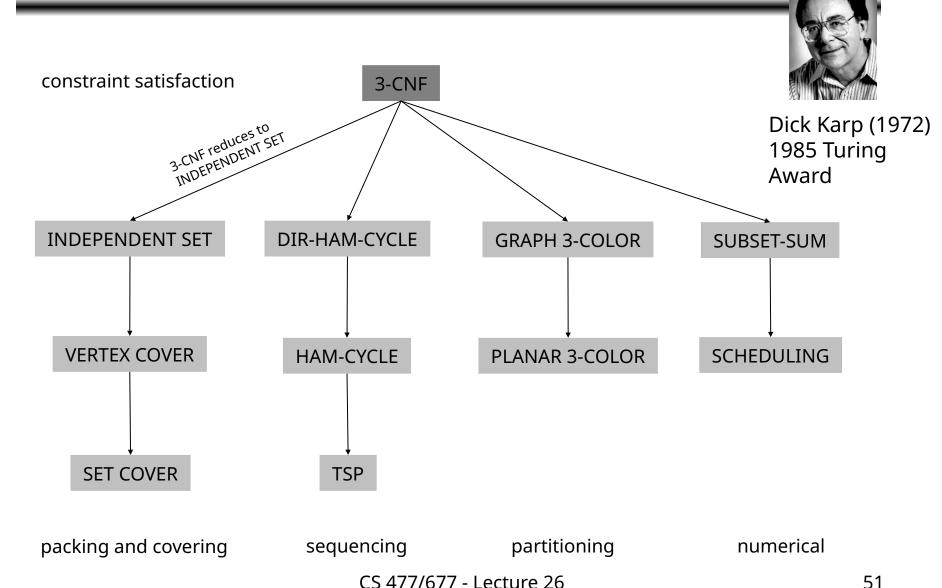
- Claim: G contains independent set of size  $k = |\Phi|$  iff  $\Phi$  is satisfiable
- Proof: "←="
  - Each triangle has a literal that evaluates to 1
  - This is an independent set S of size k
    - If there would be an edge between vertices in S, they would have to conflict



$$\Phi = \left(\overline{x_1} \ \lor \ x_2 \ \lor \ x_3\right) \land \left(x_1 \ \lor \ \overline{x_2} \ \lor \ x_3\right) \land \left(\overline{x_1} \ \lor \ x_2 \ \lor \ x_4\right)$$
CS 477/677 - Lecture 26

G

Polynomial-Time Reductions



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### Vertex Cover

- G = (V, E), undirected graph
- Vertex cover = a subset V' ⊆ V
   which covers all the edges
  - if (u, v) ∈ E then u ∈ V' or v ∈ V' or both.
- **Size** of a vertex cover = number of vertices in it

#### **Problem:**

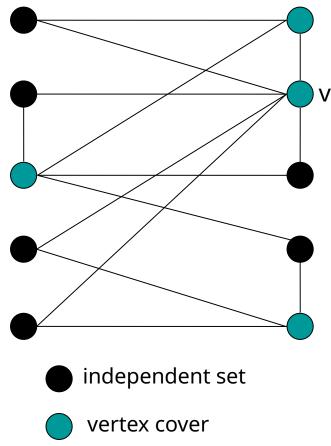
- Find a vertex cover of minimum size
- Does graph G have a vertex cover of size k?

### INDEPENDENT-SET $\leq_p$ VERTEX-COVER

 We show S is an independent set iff V ← S is a vertex cover

#### Proof "⇒"

- Let S be any independent set
- Consider an arbitrary edge (u, v)
- S independent ⇒  $u \notin S$  or  $v \notin S$ ⇒  $u \in V - S$  or  $v \in V - S$
- Thus, V S covers (u, v)

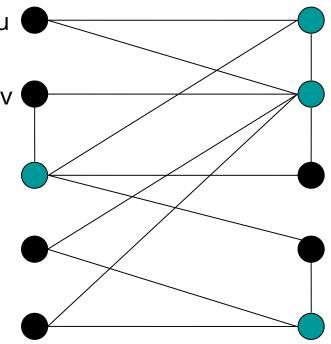


### INDEPENDENT-SET $\leq_p$ VERTEX-COVER

 We show S is an independent set iff V ← S is a vertex cover

#### Proof "←="

- Let V S be any vertex cover
- Consider two nodes u ∈ S and v ∈ S
- Observe that (u, v) ∉ E since
   V S is a vertex cover
- Thus, no two nodes in S are joined
   by an edge ⇒ S independent set



- independent set
- vertex cover

### Set Cover

• Given a set U of elements, a collection  $S_1, S_2, ...$ .,  $S_m$  of subsets of U, and an integer k, does there exist a collection of  $\leq$  k of these sets whose union is equal to U?

Example

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

k = 2

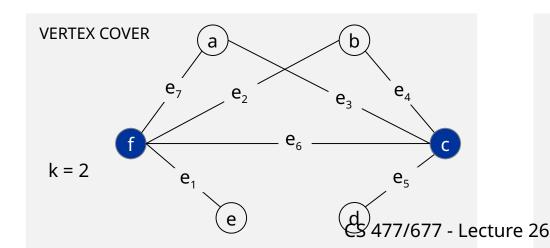
$$S_1 = \{3, 7\}$$
 $S_4 = \{2, 4\}$ 
 $S_2 = \{3, 4, 5, 6\}$ 
 $S_5 = \{5\}$ 
 $S_3 = \{1\}$ 
 $CS 477/677 - Lecture 26$ 

### Set Cover

- Given a set U of elements, a collection  $S_1, S_2, ...$ .,  $S_m$  of subsets of U, and an integer k, does there exist a collection of  $\leq$  k of these sets whose union is equal to U?
- Sample application
  - m available pieces of software
  - Set U of n capabilities that the system should 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

# $VERTEX-COVER \leq_p SET-COVER$

- Given a VERTEX-COVER instance G = (V, E), k, we construct a set cover instance whose size equals the size of the vertex cover instance
- Construction
  - Create SET-COVER instance
    - k = k, U = E,  $S_v = \{e \in E : e \text{ incident to } v \}$
  - Set-cover of size ≤ k iff vertex cover of size ≤ k

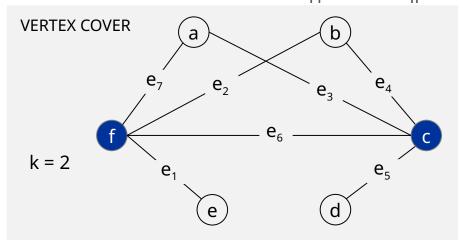


```
SET COVER

U = \{ 1, 2, 3, 4, 5, 6, 7 \}
k = 2
S_a = \{3, 7\}
S_c = \{3, 4, 5, 6\}
S_d = \{5\}
S_e = \{1\}
S_f = \{1, 2, 6, 7\}
```

# $VERTEX-COVER \leq_p SET-COVER$

- Set-cover of size  $\leq$  k iff vertex cover of size  $\leq$  k
- Proof " $\Rightarrow$ " ( $S_{i1}$ , ....,  $S_{il}$  are  $l \le k$  sets that cover U)
  - Every edge in G is incident on one of the vertices  $i_1,...,i_l$ , so  $\{i_1,...,i_l\}$  is a vertex cover of size  $l \le k$
- Proof " $\Leftarrow$ "  $\{i_1,...,i_l\}$  is a vertex cover of size  $l \leq k$ 
  - Then, the sets  $S_{i1}$ , ....,  $S_{il}$  cover U



SET COVER

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

$$k = 2$$

$$S_a = \{3, 7\}$$

$$S_b = \{2, 4\}$$

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

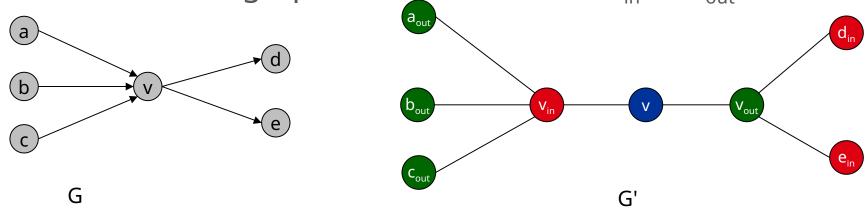
$$S_d = \{5\}$$

$$S_e = \{1\}$$

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

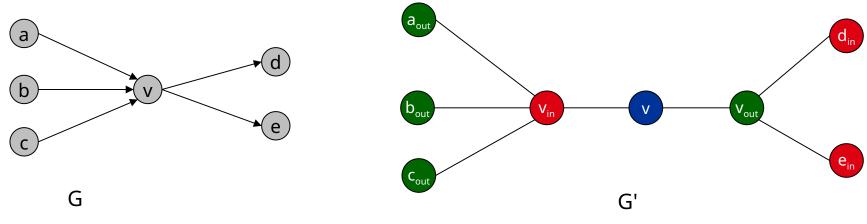
### Hamiltonian Cycle

- Given an undirected graph G = (V, E), does there exists a simple directed cycle  $\Gamma$  that contains every node in V?
- Claim: DIR-HAM-CYCLE ≤<sub>P</sub> HAM-CYCLE
- Construction
  - Given a directed graph G = (V, E), construct an undirected graph G' with 3n nodes:  $v_{in}$ , v,  $v_{out}$



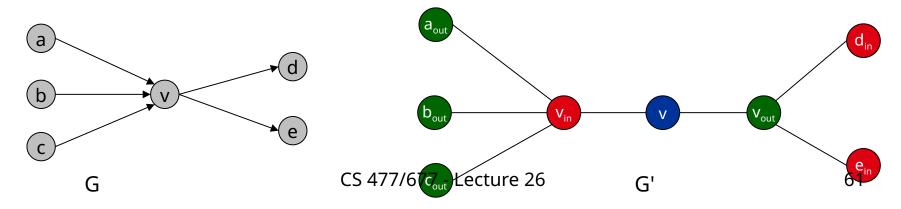
# DIR-HAM-CYCLE ≤<sub>p</sub> HAM-CYCLE

- Claim: G has a Hamiltonian cycle iff G' does.
- Proof: "⇒"
  - Suppose G has a directed Hamiltonian cycle  $\Gamma$
  - Then G' has an undirected Hamiltonian cycle (same order)



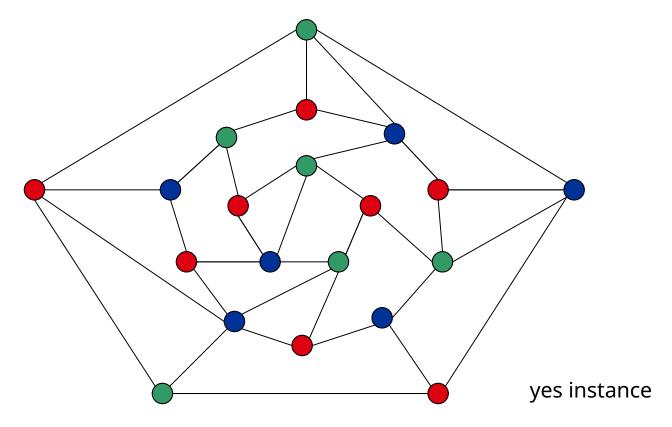
### DIR-HAM-CYCLE ≤<sub>P</sub> HAM-CYCLE

- Claim: G has a Hamiltonian cycle iff G' does.
- Proof: "←="
  - Suppose G' has an undirected Hamiltonian cycle  $\Gamma'$
  - $\Gamma$ ' must visit nodes in G' using one of following two orders:
    - ..., B, G, R, B, G, R, B, G, R, B, ...
    - ..., B, R, G, B, R, G, B, R, G, B, ...
  - Blue nodes in  $\Gamma$ ' make up directed Hamiltonian cycle  $\Gamma$  in G, or reverse of one



### 3-Colorability

 Given an undirected graph G does there exists a way to color the nodes red, green, and blue so that no adjacent nodes have the same color?



### Register Allocation

#### Register allocation

 Assign program variables to machine register so that no more than k registers are used and no two program variables that are needed at the same time are assigned to the same register

#### Interference graph

 Nodes are program variables names, edge between u and v if there exists an operation where both u and v are "live" at the same time.

#### Observation [Chaitin 1982]

 Can solve register allocation problem iff interference graph is k-colorable

#### Fact

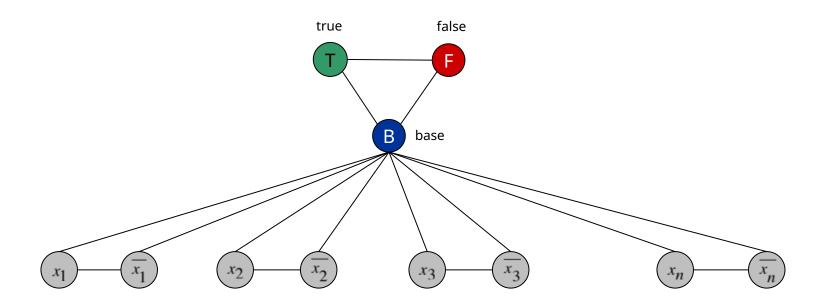
- 3-COLOR ≤  $_{P}$  k-REGISTER-ALLOCATION for any constant k ≥ 3

• Given 3-CNF instance  $\Phi$ , we construct an instance of 3-COLOR that is 3-colorable iff  $\Phi$  is satisfiable

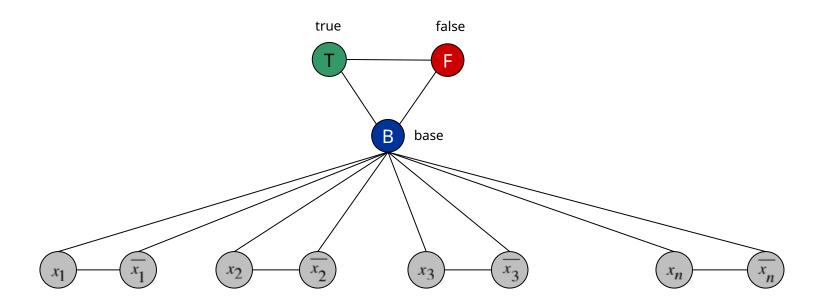
#### Construction

- For each literal, create a node
- Create 3 new nodes T, F, B; connect them in a triangle, and connect each literal to B
- Connect each literal to its negation
- For each clause, add a 6-node subgraph

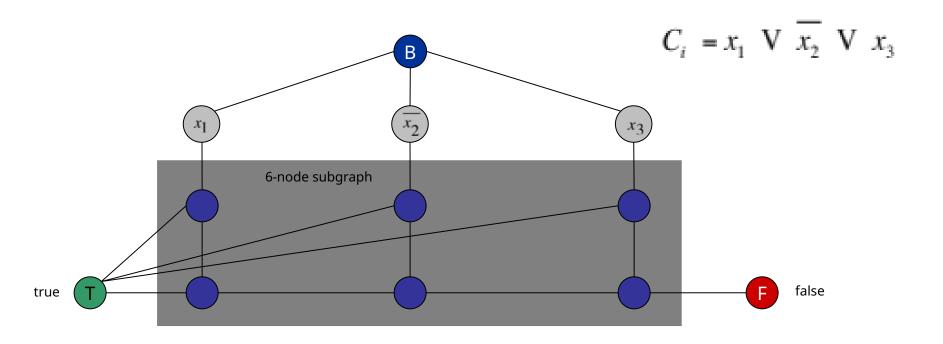
- For each literal, create a node
- Create 3 new nodes T, F, B; connect them in a triangle, and connect each literal to B
- Connect each literal to its negation



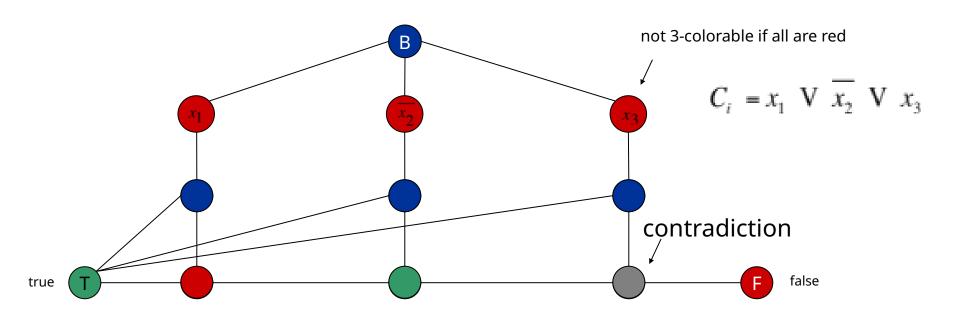
- Any 3-coloring implicitly determines a truth assignment for variables in 3-CNF
  - Nodes T, F, B must get different colors
  - For x<sub>i</sub> and not-x<sub>i</sub> one will take T color one F color



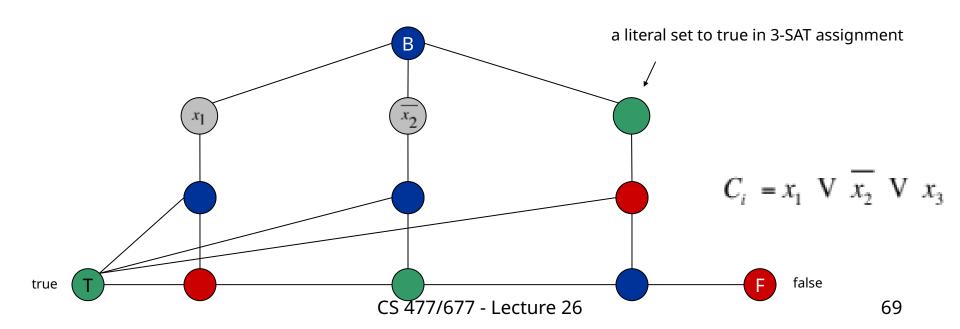
- Must ensure that only satisfying assignments can result in 3-coloring of the full graph
  - For each clause, add a 6-node subgraph



- Proof "⇒" Suppose graph is 3-colorable
  - Proof by contradiction: assume that all three literals get a False color



- Proof " $\longleftarrow$ " Suppose 3-CNF formula  $\Phi$  is satisfiable
  - Color all true literals T
  - Color node below green node F, and node below B
  - Color remaining middle row nodes B
  - Color remaining bottom nodes T or F as forced



### Directed Hamiltonian Cycle

• Given a digraph G = (V, E), does there exists a simple directed cycle  $\Gamma$  that contains every node in V?

#### Idea:

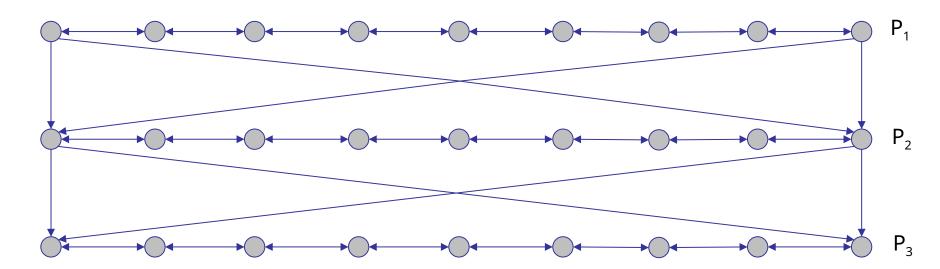
– Given an instance  $\Phi$  of 3-CNF, we construct an instance of DIR-HAM-CYCLE that has a Hamiltonian cycle iff  $\Phi$  is satisfiable

#### Construction

 Create a graph that has 2<sup>n</sup> Hamiltonian cycles which correspond in a natural way to 2<sup>n</sup> possible truth assignments

## $3-CNF \leq_p DIR-HAM-CYCLE$

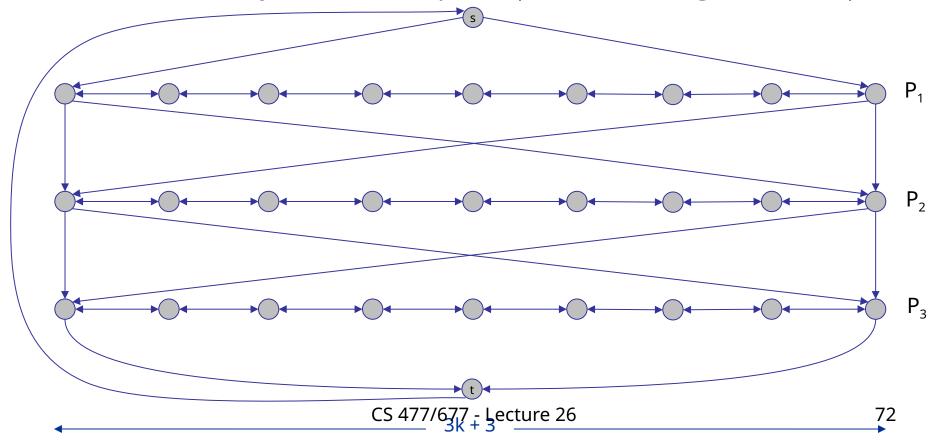
- Construction: given 3-CNF instance 
   Φ with n
   variables x<sub>i</sub> and k clauses C<sub>1</sub>, ..., C<sub>k</sub>
  - Construct n paths  $P_1$ , ...,  $P_n$ , with  $P_i$  containing  $v_{i1}$ ,  $v_{i2}$ ...,  $v_{ib}$
  - There are edges between adjacent vertices on path in each direction
  - Hook the paths together with edges



## 3-CNF ≤<sub>p</sub> DIR-HAM-CYCLE

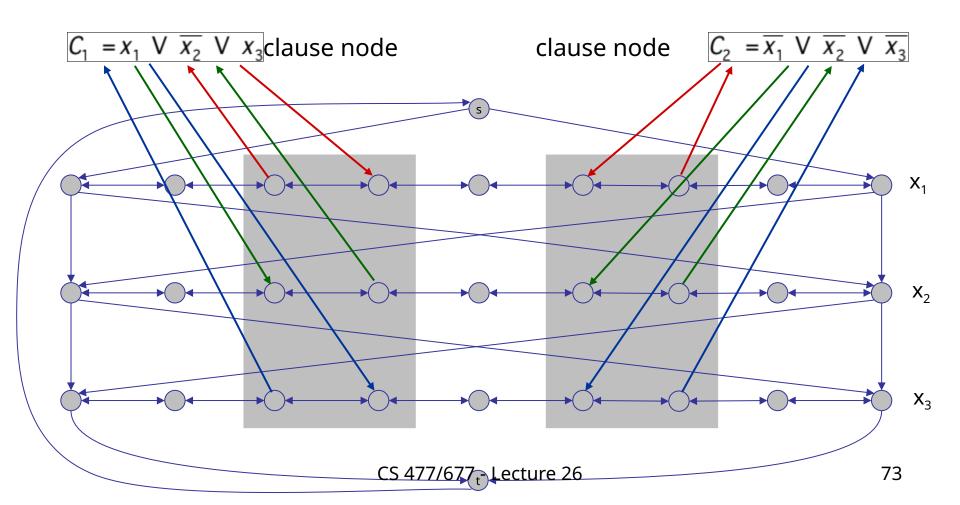
#### Construction (continued)

- Add two vertices s and t and connect them with edges
- Add edge from t to s
- Intuition: cycle traverses path  $P_i$  from left to right  $\Leftrightarrow$  set  $x_i = 1$



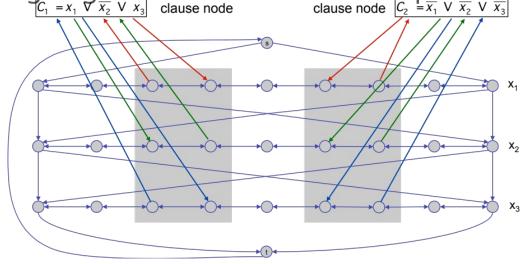
## 3-CNF ≤<sub>p</sub> DIR-HAM-CYCLE

- Construction (continued)
  - For each clause: add a node and 6 edges



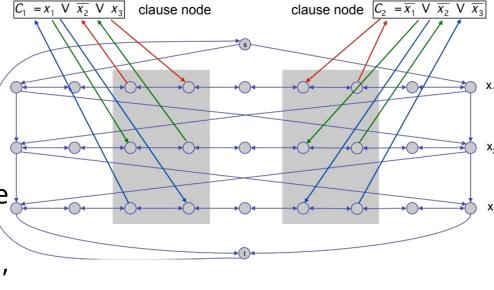
## $3-CNF \leq_p DIR-HAM-CYCLE$

- Claim:  $\Phi$  is satisfiable iff G has a Hamiltonian cycle
- Proof "⇒" Suppose 3-CNF has satisfying assignment x\*
  - Then, define Hamiltonian cycle in G as follows:
    - If  $x_i^* = 1$ , traverse row i from left to right
    - If  $x_i^* = 0$ , traverse row i from right to left
    - For each clause  $C_j$ , there will be at least one row i in which we are going in "correct" direction to splice node  $C_j$  into tour clause node  $C_j = x_1 \vee x_2 \vee x_3$  clause node  $C_j = x_1 \vee x_2 \vee x_3$



# 3-CNF ≤<sub>p</sub> DIR-HAM-CYCLE

- Claim: **Φ** is satisfiable iff G has a Hamiltonian cycle
- Proof " $\Leftarrow$ " Suppose G has a Hamiltonian cycle  $\Gamma$ 
  - If  $\Gamma$  enters clause node  $C_i$ , it must depart on mate edge
    - Nodes before and after C<sub>j</sub> are connected by an edge e in G
    - Removing C<sub>j</sub> from cycle, replace it with edge e ⇒ Hamiltonian cycle on G { C<sub>i</sub> }
  - Continuing in this way, ⇒
     Hamiltonian cycle Γ' in
     G { C<sub>1</sub> , C<sub>2</sub> , ..., C<sub>k</sub> }
  - Set  $x_i^* = 1$  iff  $\Gamma$ ' traverses row i left to right, otherwise set to 0
  - Since Γ visits each clause node C<sub>j</sub>, at least one of the paths is traversed in "correct" direction, and each clause is satisfied

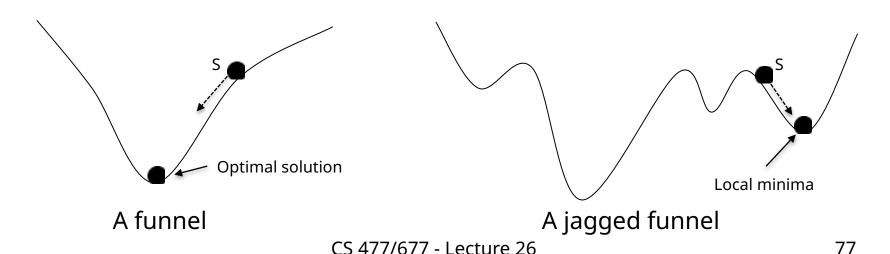


Optional, not required for final exam

### ADDITIONAL APPROXIMATION ALGORITHMS

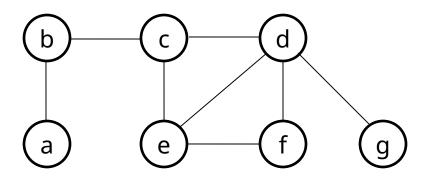
### Local Search (Hill Climbing, Gradient Descent)

- Explore the space of possible solutions, moving from a current solution to a "nearby" one
  - 1. Let S denote current solution
  - 2. If there is a neighbor S' of S with strictly lower cost, replace S with the neighbor whose cost is as small as possible
  - 3. Otherwise, terminate the algorithm



### The Vertex-Cover Problem

- Vertex cover of G = (V, E), undirected graph
  - A subset V' ⊆ V thatcovers all the edges in G

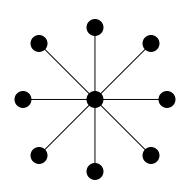


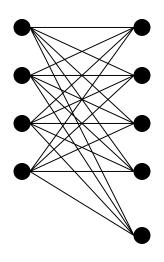
### Hill climbing (gradient descent) idea:

- Start with a solution S = V
- If there is a neighbor S' that is a vertex cover and has lower cardinality, replace S with S'.
- Algorithm ends after at most n steps (each update decreases the size of the cover by one)

### Gradient Descent: Vertex Cover

Local optimum. No neighbor is strictly better.





optimum = center node only local optimum = all other nodes

optimum = all nodes on left side local optimum = all nodes on right side



optimum = even nodes local optimum = omit every third node

### The Set Covering Problem

- Finite set X
- Family F of subsets of X:  $F = \{S_1, S_2, ..., S_n\}$

$$X = \bigcup_{S \in F} S$$

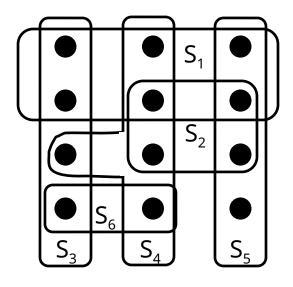
- Find a minimum-size subset C ⊆ F that covers all the elements in X
- Decision: given a number k find if there exist k sets  $S_{i1}$ ,  $S_{i2}$ , ...,  $S_{ik}$  such that:

$$S_{i1} \bigcup S_{i2} \bigcup ... \bigcup S_{ik} = X$$

### **Greedy Set Covering**

#### Idea:

At each step pick a set S
 that covers the greatest
 number of remaining
 elements



Optimal:  $C = \{S_3, S_4, S_5\}$ 

### GREEDY-SET-COVER(X, F)

- 1. U ← X
- 2. C ← Ø
- **3.** while ∪ **=**/∅
- 4. **do** select an  $S \in F$  that maximizes  $|S \cap U|$
- 5.  $U \leftarrow U S$
- 6.  $C \leftarrow C \cup \{S\}$
- 7. return C

